

Notes for *Foundations of Modern Analysis* by  
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# Chapter 1 – Measure Theory

## Section 1.1 – Rings and Algebras

### Problems

#### 1.1.1

$$\left(\varliminf_n E_n\right)^c = \overline{\varliminf_n E_n^c}, \quad \left(\overline{\varliminf_n E_n}\right)^c = \varliminf_n E_n^c.$$

*Solution.* Note that

$$\begin{aligned} x \in \varliminf_n E_n &\iff x \in E_n \text{ for all but finitely many } n \\ &\iff x \in E_n^c \text{ for only finitely many } n. \end{aligned}$$

Hence

$$\begin{aligned} x \in \left(\varliminf_n E_n\right)^c &\iff x \in E_n^c \text{ for infinitely many } n \\ &\iff x \in \overline{\varliminf_n E_n^c}, \end{aligned}$$

proving the first identity.

Next, let  $F_n = E_n^c$  for every  $n$ . Then

$$\overline{\varliminf_n E_n} = \overline{\varliminf_n F_n^c} = \left(\varliminf_n F_n\right)^c = \left(\varliminf_n E_n^c\right)^c$$

by the first identity, and the second identity follows.

#### 1.1.2

$$\overline{\varliminf_n E_n} = \bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} E_n, \quad \varliminf_n E_n = \bigcup_{k=1}^{\infty} \bigcap_{n=k}^{\infty} E_n.$$

*Solution.* Suppose  $x \in \overline{\lim_n E_n}$ . Then  $x \in E_n$  for infinitely many  $n$ . It follows that  $x \in \bigcup_{n=k}^{\infty} E_n$  for all  $k \in \mathbb{N}$ , and hence that  $x \in \bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} E_n$ .

Conversely, assume that  $x \in \bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} E_n$ . Then  $x \in \bigcup_{n=k}^{\infty} E_n$  for all  $k \in \mathbb{N}$ . It follows that  $x \in E_n$  for infinitely many  $n$ , and thus that  $x \in \overline{\lim_n E_n}$ . This proves the first identity.

Next, suppose that  $x \in \underline{\lim_n E_n}$ . Then  $x \in E_n$  for all but finitely many  $n$ , so there is some  $k' \in \mathbb{N}$  such that  $x \in E_n$  for all  $n \geq k'$ . It follows that  $x \in \bigcap_{n=k'}^{\infty} E_n$ , and hence that  $x \in \bigcup_{k=1}^{\infty} \bigcap_{n=k}^{\infty} E_n$ .

Conversely, assume that  $x \in \bigcup_{k=1}^{\infty} \bigcap_{n=k}^{\infty} E_n$ . Then  $x \in \bigcap_{n=k'}^{\infty} E_n$  for some  $k' \in \mathbb{N}$ , which means that  $x \in E_n$  for all  $n \geq k'$ . It follows that  $x \in E_n$  for all but finitely many  $n$ ; that is,  $x \in \underline{\lim_n E_n}$ .

### 1.1.3

If  $\mathcal{R}$  is a  $\sigma$ -ring and  $E_n \in \mathcal{R}$ , then

$$\bigcap_{n=1}^{\infty} E_n \in \mathcal{R}, \quad \overline{\lim_n E_n} \in \mathcal{R}, \quad \underline{\lim_n E_n} \in \mathcal{R}.$$

*Solution.* Note that

$$\bigcap_{n=1}^{\infty} E_n = E_1 \cap \left( \bigcap_{n=1}^{\infty} E_n \right) = E_1 - \left( E_1 - \bigcap_{n=1}^{\infty} E_n \right),$$

and that

$$E_1 - \bigcap_{n=1}^{\infty} E_n = \bigcup_{n=1}^{\infty} (E_1 - E_n) \in \mathcal{R},$$

using one of De Morgan's laws along with properties (b) and (e). It follows (by (b) again) that  $\bigcap_{n=1}^{\infty} E_n \in \mathcal{R}$ . That is,  $\mathcal{R}$  is closed under countable intersections.

Given  $k \in \mathbb{N}$ , let  $A_n = \emptyset$  for  $n < k$ , and let  $A_n = E_n$  for  $n \geq k$ . Then  $A_n \in \mathcal{R}$  for all  $n$  by (a), hence

$$\bigcup_{n=k}^{\infty} E_n = \bigcup_{n=1}^{\infty} A_n \in \mathcal{R}$$

by (e). Since  $\mathcal{R}$  is closed under countable intersections, we then have

$$\overline{\lim_n E_n} = \bigcap_{k=1}^{\infty} \bigcup_{n=k}^{\infty} E_n \in \mathcal{R}.$$

By a similar argument we find that

$$\bigcap_{n=k}^{\infty} E_n \in \mathcal{R}$$

for all  $k \in \mathbb{N}$ . Thus

$$\varliminf_n E_n = \bigcup_{k=1}^{\infty} \bigcap_{n=k}^{\infty} E_n \in \mathcal{R}$$

by (e).

#### 1.1.4

The intersection of any collection of rings (algebras,  $\sigma$ -rings, or  $\sigma$ -algebras) is also a ring (an algebra,  $\sigma$ -ring, or  $\sigma$ -algebra).

*Solution.* Let  $\mathcal{C}$  be a collection of classes. Let  $\bigcap \mathcal{C}$  denote the intersection of all classes in  $\mathcal{C}$ . We will show that if one of the properties (a)-(e) is satisfied by all classes in  $\mathcal{C}$ , then  $\bigcap \mathcal{C}$  satisfies that property as well. The result requested in the problem then follows as an immediate corollary.

It is clear that if every  $\mathcal{R} \in \mathcal{C}$  satisfies (a), then so does  $\bigcap \mathcal{C}$ . Suppose every  $\mathcal{R} \in \mathcal{C}$  satisfies (b). If  $A, B \in \bigcap \mathcal{C}$  then  $A, B \in \mathcal{R}$  for every  $\mathcal{R} \in \mathcal{C}$ . Hence  $A - B \in \mathcal{R}$  for all  $\mathcal{R} \in \mathcal{C}$ , and it follows that  $A - B \in \bigcap \mathcal{C}$ . The argument for (c) is similar (with  $A \cup B$  in place of  $A - B$ ), and (d) is obvious.

Finally, suppose that every  $\mathcal{R} \in \mathcal{C}$  satisfies (e). If  $A_1, A_2, \dots \in \bigcap \mathcal{C}$  then  $A_1, A_2, \dots \in \mathcal{R}$  for every  $\mathcal{R} \in \mathcal{C}$ . Hence  $\bigcup_{n=1}^{\infty} A_n \in \mathcal{R}$  for all  $\mathcal{R} \in \mathcal{C}$ , and it follows that  $\bigcup_{n=1}^{\infty} A_n \in \bigcap \mathcal{C}$ .

#### 1.1.5

If  $\mathcal{D}$  is any class of sets, then there exists a unique ring  $\mathcal{R}_0$  such that (i)  $\mathcal{R}_0 \supset \mathcal{D}$ , and (ii) any ring  $\mathcal{R}$  containing  $\mathcal{D}$  contains also  $\mathcal{R}_0$ .  $\mathcal{R}_0$  is called the *ring generated by  $\mathcal{D}$* , and is denoted by  $\mathcal{R}(\mathcal{D})$ .

*Solution.* Let  $\mathcal{R}_0$  be the intersection of all rings containing  $\mathcal{D}$ . This is a ring by the previous exercise, and it satisfies the properties (i) and (ii). To see that it is unique, let  $\mathcal{R}'_0$  also be a ring satisfying (i) and (ii). Then  $\mathcal{R}_0 \subset \mathcal{R}'_0$  and  $\mathcal{R}'_0 \subset \mathcal{R}_0$  by property (ii), so  $\mathcal{R}_0 = \mathcal{R}'_0$ .

#### 1.1.6

If  $\mathcal{D}$  is any class of sets, then there exists a unique  $\sigma$ -ring  $\mathcal{S}_0$  such that (i)  $\mathcal{S}_0 \supset \mathcal{D}$ , and (ii) any  $\sigma$ -ring containing  $\mathcal{D}$  contains also  $\mathcal{S}_0$ . We call  $\mathcal{S}_0$  the  *$\sigma$ -ring generated by  $\mathcal{D}$* , and denote it by  $\mathcal{S}(\mathcal{D})$ . A similar result holds for  $\sigma$ -algebras, and we speak of the  *$\sigma$ -algebra generated by  $\mathcal{D}$* .

*Solution.* By the same argument as in the previous exercise,  $\mathcal{S}_0$  is the intersection of all  $\sigma$ -rings containing  $\mathcal{D}$ . Similarly the  $\sigma$ -algebra generated by  $\mathcal{D}$  is the intersection of all  $\sigma$ -algebras containing  $\mathcal{D}$ .

#### 1.1.7

If  $\mathcal{D}$  is any class of sets, then every set in  $\mathcal{R}(\mathcal{D})$  can be covered by (that is, is contained in) a finite union of sets of  $\mathcal{D}$ . [*Hint:* The class  $\mathcal{K}$  of sets that can be covered by finite unions of sets of  $\mathcal{D}$  forms a ring.]

*Solution.* Let  $\mathcal{K}$  be the class of all sets that can be covered by a finite union of sets in  $\mathcal{D}$ . Certainly  $\emptyset \in \mathcal{K}$ , since  $\emptyset$  is a subset of the empty union. If  $A, B \in \mathcal{K}$ , then

$$A \subset \bigcup_{i=1}^m E_i, \quad B \subset \bigcup_{i=1}^n F_i,$$

for some sets  $E_1, \dots, E_m, F_1, \dots, F_n \in \mathcal{D}$ . (Note that  $m$  or  $n$  can be zero, in which case the corresponding union is empty.) Thus

$$A - B \subset A \subset \bigcup_{i=1}^m E_i$$

and

$$A \cup B \subset \left( \bigcup_{i=1}^m E_i \right) \cup \left( \bigcup_{j=1}^n F_j \right),$$

so both  $A - B$  and  $A \cup B$  are elements of  $\mathcal{K}$ .

The above shows that  $\mathcal{K}$  is a ring, and certainly  $\mathcal{D} \subset \mathcal{K}$ . Hence  $\mathcal{R}(\mathcal{D}) \subset \mathcal{K}$  by Problem 1.1.5, and it follows that every set in  $\mathcal{R}(\mathcal{D})$  can be covered by a finite union of sets in  $\mathcal{D}$ .

## Section 1.2 – Definition of Measure

### Problems

#### 1.2.1

If  $\mu$  satisfies the properties (i)-(iii) in Definition 1.2.1, and if  $\mu(E) < \infty$  for at least one set  $E$ , then (iv) is also satisfied.

*Solution.* We have

$$\mu(E) = \mu(E \cup \emptyset) = \mu(E) + \mu(\emptyset),$$

hence  $\mu(\emptyset) = 0$ .

#### 1.2.2

Let  $X$  be an infinite space. Let  $\mathcal{A}$  be the class of all subsets of  $X$ . Define  $\mu(E) = 0$  if  $E$  is finite and  $\mu(E) = \infty$  if  $E$  is infinite. Then  $\mu$  is finitely additive but not completely additive.

*Solution.* Suppose  $A, B \in \mathcal{A}$ . Note that  $A \cup B$  is finite if both  $A$  and  $B$  are finite, but infinite otherwise. Hence

$$\mu(A \cup B) = 0 = \mu(A) + \mu(B)$$

in the former case, and

$$\mu(A \cup B) = \infty = \mu(A) + \mu(B)$$

in the latter. This proves that  $\mu$  is additive; *finite* additivity follows by a simple induction argument.

Let  $(x_n)$  be a sequence of distinct points in  $X$ . Then  $\bigcup_{n=1}^{\infty} \{x_n\}$  is an infinite set, so

$$\mu\left(\bigcup_{n=1}^{\infty} \{x_n\}\right) = \infty,$$

but

$$\sum_{n=1}^{\infty} \mu(\{x_n\}) = 0.$$

Thus  $\mu$  is not completely additive.

### 1.2.3

If  $\mu$  is a measure on a  $\sigma$ -algebra  $\mathcal{A}$ , and if  $E, F$  are sets of  $\mathcal{A}$ , then

$$\mu(E) + \mu(F) = \mu(E \cup F) + \mu(E \cap F).$$

*Solution.* If  $\mu(F) = \infty$ , then  $\mu(E \cup F) = \infty$  by Theorem 1.2.1(i), and the given equality holds. If  $\mu(F) < \infty$ , then

$$\begin{aligned} \mu(E \cup F) &= \mu[E \cup (F - (E \cap F))] \\ &= \mu(E) + \mu[F - (E \cap F)] \\ &= \mu(E) + \mu(F) - \mu(E \cap F), \end{aligned}$$

with the last equality following from Theorem 1.2.1(ii). Note that  $E \cap F \subset F$  so that  $\mu(E \cap F) \leq \mu(F) < \infty$ . Hence we can rearrange the above to yield

$$\mu(E) + \mu(F) = \mu(E \cup F) + \mu(E \cap F).$$

### 1.2.6

Give an example of a measure  $\mu$  and a monotone-decreasing sequence  $\{E_n\}$  of  $\mathcal{A}$  such that  $\mu(E_n) = \infty$  for all  $n$ , and  $\mu(\lim_n E_n) = 0$ .

*Solution.* Let  $X = \mathbb{R}$  and let  $\mathcal{A} = \mathcal{P}(\mathbb{R})$  (the power set of  $\mathbb{R}$ ; this is easily seen to be a  $\sigma$ -algebra). Define  $\mu : \mathcal{A} \rightarrow [0, \infty]$  such that  $\mu(E)$  is the number of points in  $E$  (with  $\mu(E) = \infty$  if  $E$  is infinite). This is easily seen to be a measure.

For each  $n \in \mathbb{N}$ , let  $E_n = (0, 1/n)$ . Then  $(E_n)$  is a monotone decreasing sequence of sets in  $\mathcal{A}$ ,  $\mu(E_n) = \infty$  for all  $n$ , and

$$\mu\left(\lim_n E_n\right) = \mu\left(\bigcap_{n=1}^{\infty} E_n\right) = \mu(\emptyset) = 0.$$

## Section 1.3 – Outer Measure

### Problems

#### 1.3.1

Define  $\mu^*(E)$  as the number of points in  $E$  if  $E$  is finite and  $\mu^*(E) = \infty$  if  $E$  is infinite. Show that  $\mu^*$  is an outer measure. Determine the measurable sets.

*Solution.* Of the properties listed in Definition 1.3.1, only countable subadditivity is non-obvious for  $\mu^*$ . But let us start with proving finite subadditivity.

Let  $A$  and  $B$  be sets. If either is infinite, then so is  $A \cup B$ , hence

$$\mu^*(A \cup B) = \infty = \mu^*(A) + \mu^*(B).$$

If both  $A$  and  $B$  are finite sets, then

$$\mu^*(A \cup B) = \mu^*(A) + \mu^*(B - A) \leq \mu^*(A) + \mu^*(B)$$

by basic set-theoretic considerations. Thus  $\mu^*$  is subadditive, and finite subadditivity follows by induction on the number of sets in the union.

Now, let  $(E_n)$  be a sequence of sets. If infinitely many of the sets  $E_n$  are nonempty, then  $\sum_n \mu^*(E_n) = \infty$ , and

$$\mu^*\left(\bigcup_n E_n\right) \leq \sum_n \mu^*(E_n)$$

follows. If only finitely many of the sets  $E_n$  are nonempty, let  $E_{n_1}, E_{n_2}, \dots, E_{n_k}$  be those sets. Then

$$\mu^*\left(\bigcup_{n=1}^{\infty} E_n\right) = \mu^*\left(\bigcup_{i=1}^k E_{n_i}\right) \leq \sum_{i=1}^k \mu^*(E_{n_i}) = \sum_{n=1}^{\infty} \mu^*(E_n),$$

by finite subadditivity. This proves that  $\mu^*$  is countably subadditive, and hence that  $\mu^*$  is an outer measure.

Note that  $\mu^*$  is *additive* on disjoint sets; if  $A \cap B = \emptyset$ , then  $\mu^*(A \cup B) = \mu^*(A) + \mu^*(B)$ . In particular,

$$\mu^*(A) = \mu^*(A \cap E) + \mu^*(A - E)$$

for all sets  $A, E$ . That is, all sets are measurable.

#### 1.3.2

Define  $\mu^*(\emptyset) = 0$ ,  $\mu^*(E) = 1$  if  $E \neq \emptyset$ . Show that  $\mu^*$  is an outer measure, and determine the measurable sets.

*Solution.* As in the previous exercise, the only slightly non-obvious property is countable subadditivity. Hence, let  $(E_n)$  be a sequence of sets. If all the sets  $E_n$  are empty, then certainly

$$\mu^*\left(\bigcup_n E_n\right) = 0 = \sum_n \mu^*(E_n).$$

If not, then there is some  $m$  such that  $E_m \neq \emptyset$ , and it follows that

$$\mu^*\left(\bigcup_n E_n\right) = 1 = \mu^*(E_m) \leq \sum_n \mu^*(E_n).$$

Thus  $\mu^*$  is indeed countably subadditive, and therefore also an outer measure.

The empty set is measurable:

$$\mu^*(A \cap \emptyset) + \mu^*(A - \emptyset) = \mu^*(\emptyset) + \mu^*(A) = \mu^*(A)$$

for all sets  $A$ . It follows by Theorem 1.3.1 that  $X$  is measurable as well (the measurable sets make up a  $\sigma$ -algebra). Indeed  $\emptyset$  and  $X$  are the only measurable sets. To see this, let  $E$  be any set other than those two (this requires that  $X$  contains at least two elements). Then both  $E$  and  $E^c$  are nonempty, so

$$\mu^*(X \cap E) + \mu^*(X - E) = \mu^*(E) + \mu^*(E^c) = 2 > 1 = \mu^*(X).$$

## Section 1.4 – Construction of Outer Measures

### Problems

#### 1.4.4

If  $\mathcal{K}$  is a  $\sigma$ -algebra and  $\lambda$  is a measure on  $\mathcal{K}$ , then  $\mu^*(A) = \lambda(A)$  for any  $A \in \mathcal{K}$ . [*Hint:*  $\mu^*(A) = \inf\{\lambda(E); E \in \mathcal{K}, E \supset A\}$ .]

*Solution.* Note that the description of  $\mu^*$  can be simplified when  $\mathcal{K}$  is a  $\sigma$ -algebra and  $\lambda$  is a measure. For suppose that  $A \subset X$ ,  $E_n \in \mathcal{K}$  ( $n = 1, 2, \dots$ ), and  $A \subset \bigcup_n E_n$ . Then  $E := \bigcup_n E_n \in \mathcal{K}$ , and  $\lambda(E) \leq \sum_n \lambda(E_n)$  by Theorem 1.2.2. Hence

$$\mu^*(A) = \inf\{\lambda(E); E \in \mathcal{K}, E \supset A\}.$$

Now, suppose that  $A \in \mathcal{K}$ . Certainly  $\lambda(A)$  is an element of  $\{\lambda(E); E \in \mathcal{K}, E \supset A\}$ . And if  $E \in \mathcal{K}$  and  $E \supset A$ , then  $\lambda(E) \geq \lambda(A)$  by Theorem 1.2.1(i). Thus

$$\lambda(A) = \inf\{\lambda(E); E \in \mathcal{K}, E \supset A\} = \mu^*(A).$$



### 1.4.5

If  $\mathcal{K}$  is a  $\sigma$ -algebra and  $\lambda$  is a measure on  $\mathcal{K}$ , then every set in  $\mathcal{K}$  is  $\mu^*$ -measurable.

*Solution.* Recall the simplified description of  $\mu^*$  from the previous problem. Let  $E \in \mathcal{K}$  and  $A \subset X$ . For every  $\epsilon > 0$  there exists  $F \in \mathcal{K}$  such that  $F \supset A$  and

$$\mu^*(A) + \epsilon > \lambda(F);$$

else  $\mu^*(A)$  would not be the greatest lower bound of  $\{\lambda(E); E \in \mathcal{K}, E \supset A\}$ . Moreover,

$$\lambda(F) = \lambda(F \cap E) + \lambda(F - E)$$

since  $\lambda$  is a measure on  $\mathcal{K}$ ,

$$\lambda(F \cap E) + \lambda(F - E) = \mu^*(F \cap E) + \mu^*(F - E)$$

by what we found in the previous exercise, and finally

$$\mu^*(F \cap E) + \mu^*(F - E) \geq \mu^*(A \cap E) + \mu^*(A - E)$$

by monotonicity of the outer measure  $\mu^*$ . Putting all of this together, we have

$$\mu^*(A) + \epsilon > \mu^*(A \cap E) + \mu^*(A - E)$$

for all  $\epsilon > 0$ , and thus

$$\mu^*(A) \geq \mu^*(A \cap E) + \mu^*(A - E).$$

It follows that every set  $E \in \mathcal{K}$  is  $\mu^*$ -measurable.

## Section 1.6 – The Lebesgue and the Lebesgue-Stieltjes Measures

### Problems

#### 1.6.3

The outer Lebesgue measure of a closed bounded interval  $[a, b]$  on the real line is equal to  $b - a$ . [*Hint:* Use the Heine-Borel theorem to replace a countable covering by a finite covering.]

*Solution.* Suppose  $(E_n)$  is a sequence of elements of  $\mathcal{K}$  (i.e. a sequence of open intervals) such that  $[a, b] \subset \bigcup_{n=1}^{\infty} E_n$ . The collection  $\{E_n\}$  constitutes an *open cover* of  $[a, b]$ . By the Heine-Borel theorem  $[a, b]$  is compact, hence there exists a *finite subcover*  $\{E_{n_1}, \dots, E_{n_k}\}$ , such that  $[a, b] \subset \bigcup_{i=1}^k E_{n_i}$ .

Assume without loss of generality that  $E_{n_i} \cap [a, b] \neq \emptyset$  for all  $i$ ; otherwise we can simply remove those  $E_{n_i}$  that are disjoint with  $[a, b]$  and still have a finite subcover. Write  $E_{n_i} = (a_i, b_i)$  for each  $i$ , and define

$$\alpha = \min\{a_1, \dots, a_k\}, \quad \beta = \max\{b_1, \dots, b_k\}.$$

It is clear that  $\alpha$  and  $\beta$  are the infimum and supremum, respectively, of  $\bigcup_{i=1}^k E_{n_i}$ . Note that  $\alpha = a_j$  for some  $j$ , and  $a_j < a < b_j$  since  $E_{n_j}$  and  $[a, b]$  have nonempty intersection. Thus  $(\alpha, a] \subset \bigcup_{i=1}^k E_{n_i}$ , and similarly  $[b, \beta) \subset \bigcup_{i=1}^k E_{n_i}$ . It follows that

$$\bigcup_{i=1}^k E_{n_i} = (\alpha, \beta) \in \mathcal{K}.$$

Finally note that  $\lambda$  is finitely subadditive. (This is easily proven with induction.) (TODO: This is not convincing; use better proof from Rosenthal notes.) Thus,

$$\sum_{n=1}^{\infty} \lambda(E_n) \geq \sum_{i=1}^k \lambda(E_{n_i}) \geq \lambda[(\alpha, \beta)] = \beta - \alpha > b - a.$$

It follows that  $b - a$  is a lower bound of the set

$$\Lambda([a, b]) := \left\{ \sum_{n=1}^{\infty} \lambda(E_n); E_n \in \mathcal{K}, \bigcup_{n=1}^{\infty} E_n \supset [a, b] \right\}.$$

Moreover, for every  $\epsilon > 0$  we have

$$[a, b] \subset \left(a - \frac{\epsilon}{2}, b + \frac{\epsilon}{2}\right) \in \mathcal{K}$$

and

$$\lambda \left[ \left(a - \frac{\epsilon}{2}, b + \frac{\epsilon}{2}\right) \right] = b - a + \epsilon.$$

Hence  $b - a$  is the *greatest* lower bound of  $\Lambda([a, b])$ , and  $\mu^*([a, b]) = b - a$ .

#### 1.6.4

The outer Lebesgue measure of each of the intervals  $(a, b)$ ,  $[a, b)$ ,  $(a, b]$  is equal to  $b - a$ .

*Solution.* Recall that  $\mu^*$  is monotone, on account of being an outer measure. Hence  $\mu^*[(a, b)] \leq \mu^*([a, b]) = b - a$ , the latter equality being the result of the previous problem. Moreover, for all  $\epsilon \in (0, b - a)$  we have

$$\left(a + \frac{\epsilon}{2}, b - \frac{\epsilon}{2}\right) \subset (a, b),$$

so that

$$\mu^*[(a, b)] \geq \mu^* \left[ \left(a + \frac{\epsilon}{2}, b - \frac{\epsilon}{2}\right) \right] = b - a + \epsilon.$$

Thus  $\mu^*[(a, b)] \geq b - a$ , and it follows that  $\mu^*[(a, b)] = b - a$ .

The outer measures of  $[a, b)$  and  $(a, b]$  follow immediately by monotonicity:

$$\mu^*[(a, b)] \leq \mu^*([a, b)) \leq \mu^*([a, b]),$$

so that  $\mu^*([a, b)) = b - a$ . Similarly for  $(a, b]$ .

### 1.6.5

Consider the transformation  $Tx = \alpha x + \beta$  from the real line onto itself, where  $\alpha, \beta$  are real numbers and  $\alpha \neq 0$ . It maps sets  $E$  onto sets  $T(E)$ . Denote by  $\mu$  ( $\mu^*$ ) the Lebesgue measure (outer measure) on the real line. Prove

- (a) For any set  $E$ ,  $\mu^*(T(E)) = |\alpha|\mu^*(E)$ .
- (b)  $E$  is Lebesgue-measurable if and only if  $T(E)$  is Lebesgue-measurable.
- (c) If  $E$  is Lebesgue-measurable, then  $\mu(T(E)) = |\alpha|\mu(E)$ .

*Solution.* Let us start with a couple of simple observations:

- $T$  is bijective, with inverse given by

$$T^{-1}(x) = \frac{x - \beta}{\alpha}.$$

- Suppose  $I = (a, b)$ . Then

$$T(I) = (\alpha a + \beta, \alpha b + \beta)$$

if  $\alpha > 0$ , and

$$T(I) = (\beta b + \beta, \alpha a + \beta)$$

if  $\alpha < 0$ . Either way,

$$\mu^*[T(I)] = |\alpha|(b - a) = |\alpha|\mu^*(I),$$

where we have used one of the results of the previous exercise. Similarly,  $T^{-1}(I)$  is an open interval and

$$\mu^*[T^{-1}(I)] = |\alpha|^{-1}\mu^*(I).$$

Of course, the latter two identities still hold if  $I = \emptyset$ . Hence they hold for all  $I \in \mathcal{K}$ .

Also, let us use the notation

$$\Lambda(E) = \left\{ \sum_{n=1}^{\infty} \lambda(I_n); I_n \in \mathcal{K}, \bigcup_{n=1}^{\infty} I_n \supset E \right\}$$

for all  $E \subset \mathbb{R}$ .

- (a) Suppose  $(I_n)$  is a sequence in  $\mathcal{K}$  (i.e. a sequence of open intervals) and  $E \subset \bigcup_n I_n$ . Then  $T(I_n) \in \mathcal{K}$  for every  $n$ ,

$$T(E) \subset T\left(\bigcup_n I_n\right) = \bigcup_n T(I_n),$$

and

$$\sum_n \lambda[T(I_n)] = |\alpha| \sum_n \lambda(I_n).$$

Thus, if  $s \in \Lambda(E)$ , then  $|\alpha|s \in \Lambda[T(E)]$ . It follows that

$$\mu^*[T(E)] = \inf \Lambda[T(E)] \leq |\alpha| \inf \Lambda(E) = |\alpha| \mu^*(E).$$

Conversely, suppose  $(J_n)$  is a sequence in  $\mathcal{K}$  and  $T(E) \subset \bigcup_n J_n$ . Then  $T^{-1}(J_n) \in \mathcal{K}$  for all  $n$ ,

$$E = T^{-1}[T(E)] \subset T^{-1}\left(\bigcup_n J_n\right) = \bigcup_n T^{-1}(J_n),$$

and

$$\sum_n \lambda[T^{-1}(J_n)] = |\alpha|^{-1} \sum_n \lambda(J_n).$$

Hence, by the same logic as above, we find that  $\mu^*(E) \leq |\alpha|^{-1} \mu^*[T(E)]$ , and it follows that

$$\mu^*[T(E)] = |\alpha| \mu^*(E).$$

- (b) Note that if  $f : X \rightarrow Y$  is a bijective function (between arbitrary sets  $X, Y$ ), then

$$\begin{aligned} f^{-1}[f(A)] &= A, \\ f(A \cup B) &= f(A) \cup f(B), \\ f(A - B) &= f(A) - f(B), \\ f[f^{-1}(C)] &= C, \end{aligned}$$

for all  $A, B \subset X$  and  $C \subset Y$ .

Suppose that  $E$  is measurable:

$$\mu^*(A) = \mu^*(A \cap E) + \mu^*(A - E)$$

for all  $A \subset \mathbb{R}$ . Then, for all  $B \subset \mathbb{R}$ , we have

$$\begin{aligned} \mu^*[B \cap T(E)] + \mu^*[B - T(E)] &= \mu^*[T(T^{-1}(B) \cap E)] + \mu^*[T(T^{-1}(B) - E)] \\ &= |\alpha| \mu^*[T^{-1}(B) \cap E] + |\alpha| \mu^*[T^{-1}(B) - E] \\ &= |\alpha| \mu^*[T^{-1}(B)] \\ &= \mu^*(B), \end{aligned}$$

so that  $T(E)$  is measurable.

Conversely, suppose that  $T(E)$  is measurable. Then, for all  $A \subset \mathbb{R}$ ,

$$\begin{aligned} \mu^*(A \cap E) + \mu^*(A - E) &= \mu^*[T^{-1}(T(A) \cap T(E))] + \mu^*[T^{-1}(T(A) - T(E))] \\ &= |\alpha|^{-1} \mu^*[T(A) \cap T(E)] + |\alpha|^{-1} \mu^*[T(A) - T(E)] \\ &= |\alpha|^{-1} \mu^*[T(A)] \\ &= \mu^*(A), \end{aligned}$$

so that  $E$  is measurable.

- (c) This is immediate given (a), (b), and the definition of the Lebesgue-measure. First,  $T(E)$  is Lebesgue-measurable by (b). Next,  $\mu(E) = \mu^*(E)$  and  $\mu[T(E)] = \mu^*[T(E)]$  since  $\mu$  is simply the restriction of  $\mu^*$  to the measurable sets. Finally,  $\mu^*[T(E)] = |\alpha|\mu^*(E)$  by (a).

# Chapter 2 – Integration

## Section 2.1 – Definition of Measurable Functions

### Problems

#### 2.1.6

The *characteristic function* of a set  $E$  is the function  $\chi_E$  defined by

$$\chi_E(x) = \begin{cases} 1, & \text{if } x \in E, \\ 0, & \text{if } x \notin E. \end{cases}$$

Prove that the set  $E$  is measurable if and only if the function  $\chi_E$  is measurable.

*Solution.* Suppose  $E \in \mathcal{A}$ . For all  $c \in \mathbb{R}$ ,

$$\chi_E^{-1}\{(-\infty, c)\} = \{x \in X; \chi_E(x) < c\} = \begin{cases} \emptyset & (c \leq 0), \\ E^c & (0 < c \leq 1), \\ X & (c > 1), \end{cases}$$

so that  $\chi_E^{-1}\{(-\infty, c)\} \in \mathcal{A}$ . By Theorem 2.1.1,  $\chi_E$  is measurable.

Conversely, suppose  $\chi_E$  is measurable. Then  $E$  is measurable, since

$$E = X - E^c = \chi^{-1}\{(-\infty, 2)\} - \chi^{-1}\{(-\infty, 1)\}.$$

#### 2.1.9

If  $f$  is measurable, then  $|f|$  and  $|f|^2$  are measurable.

*Solution.* If  $c \leq 0$ , then

$$(|f|)^{-1}\{(-\infty, c)\} = (|f|^2)^{-1}\{(-\infty, c)\} = \emptyset \in \mathcal{A},$$

since  $|f|$  and  $|f|^2$  are nonnegative functions.

Let  $c > 0$ . Then

$$(|f|)^{-1}\{(-\infty, c)\} = \{x \in X; -c < f(x) < c\} = f^{-1}\{(-c, c)\}.$$

The set  $(-c, c)$  is open, hence  $f^{-1}\{(-c, c)\} \in \mathcal{A}$  by the measurability of  $f$ . Similarly,

$$(|f|^2)^{-1}\{(-\infty, c)\} = f^{-1}\{(-\sqrt{c}, \sqrt{c})\} \in \mathcal{A}.$$

Finally,

$$(|f|)^{-1}\{+\infty\} = (|f|^2)^{-1}\{+\infty\} = f^{-1}\{+\infty\} \cup f^{-1}\{-\infty\} \in \mathcal{A}$$

by the measurability of  $f$ , and

$$(|f|)^{-1}\{-\infty\} = (|f|^2)^{-1}\{-\infty\} = \emptyset \in \mathcal{A}$$

since  $|f|$  and  $|f|^2$  are nonnegative. Thus, both  $|f|$  and  $|f|^2$  are measurable by Theorem 2.1.1.

### 2.1.10

A monotone function defined on the real line is Lebesgue-measurable.

*Solution.* Let  $f$  be a monotone increasing extended real-valued function on  $\mathbb{R}$ ;

$$(\forall x, y \in \mathbb{R}) : \quad x < y \implies f(x) \leq f(y).$$

Given any  $c \in \mathbb{R}$ , let

$$\xi_c = \inf\{x \in X; f(x) \geq c\}.$$

We need to consider two cases:  $f(\xi_c) < c$  and  $f(\xi_c) \geq c$ . In the former case,  $f(x) < c$  for all  $x \leq \xi_c$  and  $f(x) \geq c$  for all  $x > \xi_c$  (by monotonicity). Hence

$$f^{-1}\{(-\infty, c)\} = (-\infty, \xi_c].$$

This is a Borel set, hence also a Lebesgue set (see Problem 1.9.3). In the latter case,  $f(x) < c$  for all  $x < \xi_c$  and  $f(x) \geq c$  for all  $x \geq \xi_c$ , so that

$$f^{-1}\{(-\infty, c)\} = (-\infty, \xi_c),$$

which is Lebesgue-measurable. Since  $c$  was arbitrary, we conclude that  $f$  is measurable, by Theorem 2.1.1.

The proof for  $f$  monotone decreasing is similar.

## Section 2.2 – Operations on Measurable Functions

### Problems

#### 2.2.2

Let  $g(u_1, \dots, u_k)$  be a continuous function in  $\mathbb{R}^k$ , and let  $\varphi_1, \dots, \varphi_k$  be measurable functions. Prove that the composite function  $h(x) = g[\varphi_1(x), \dots, \varphi_k(x)]$  is a measurable function. Note that as a special case we may conclude that

$$\max(\varphi, \dots, \varphi_n) \quad \text{and} \quad \min(\varphi, \dots, \varphi_n)$$

are measurable functions.

*Solution.* We will use the following fact, which may be proven in a course in topology:

$\mathbb{R}^k$  has a countable basis of product open subsets. Hence, if  $U$  is an open subset of  $\mathbb{R}^k$ , then there are open subsets  $U_{ni} \subset \mathbb{R}$  for  $n = 1, 2, \dots$  and  $i = 1, \dots, k$  such that

$$U = \bigcup_{n=1}^{\infty} (U_{n1} \times \cdots \times U_{nk}).$$

We are assuming that  $g$  is real-valued, likewise for the functions  $\varphi_i$ . Let  $c \in \mathbb{R}$ . Note that  $g^{-1}\{(-\infty, c)\}$  is open by continuity of  $g$ . Thus

$$g^{-1}\{(-\infty, c)\} = \bigcup_{n=1}^{\infty} (U_{n1} \times \cdots \times U_{nk})$$

for some open subsets  $U_{ni} \subset \mathbb{R}$ . Hence

$$\begin{aligned} h^{-1}\{(-\infty, c)\} &= \{x \in X; g(\varphi_1(x), \dots, \varphi_k(x)) \leq c\} \\ &= \{x \in X; (\varphi_1(x), \dots, \varphi_k(x)) \in g^{-1}\{(-\infty, c)\}\} \\ &= \bigcup_{n=1}^{\infty} \{x \in X; (\varphi_1(x), \dots, \varphi_k(x)) \in U_{n1} \times \cdots \times U_{nk}\} \\ &= \bigcup_{n=1}^{\infty} \bigcap_{i=1}^k \{x \in X; \varphi_i(x) \in U_{ni}\} \\ &= \bigcup_{n=1}^{\infty} \bigcap_{i=1}^k \varphi_i^{-1}(U_{ni}). \end{aligned}$$

The sets  $\varphi_i(U_{ni})$  are measurable since the functions  $\varphi_i$  are measurable. It follows that  $h^{-1}\{(-\infty, c)\}$  is measurable, and thus that  $h$  is measurable, by Theorem 2.1.1.

To apply the above to the max and min functions  $\mathbb{R}^k \rightarrow \mathbb{R}$  we must show that they are continuous. Let  $a < b$  and note that

$$\begin{aligned} \max^{-1}\{(a, b)\} &= \{(x_1, \dots, x_k) \in \mathbb{R}^k; x_i > a \text{ for some } i\} \\ &\cap \{(x_1, \dots, x_k) \in \mathbb{R}^k; x_i < b \text{ for all } i\}. \end{aligned}$$

Both sets in the above binary intersection are easily seen to be open by considering  $\epsilon$ -neighborhoods about their points. It follows that  $\max^{-1}(U)$  is open for all open subsets  $U \in \mathbb{R}^k$ , since every such  $U$  can be written as a countable union of open intervals. Thus max is continuous, and one similarly shows that min is continuous.



### 2.2.3

Let  $f(x)$  be a measurable function and define

$$g(x) = \begin{cases} \frac{1}{f(x)}, & \text{if } f(x) \neq 0, \\ 0, & \text{if } f(x) = 0. \end{cases}$$

Prove that  $g$  is measurable.

*Solution.* For  $c < 0$ ,

$$g^{-1}\{(-\infty, c)\} = \{x; 1/f(x) < c\} = \{x; 1/c < f < 0\} = f^{-1}\{(1/c, 0)\},$$

which is measurable by the measurability of  $f$ . Next,

$$g^{-1}\{(-\infty, 0)\} = \{x; 1/f(x) < 0\} = \{x; f(x) < 0\} = f^{-1}\{(-\infty, 0)\},$$

also measurable. Note that if we take the natural convention (unfortunately not addressed in the text) that  $x/(\pm\infty) = 0$  for all  $x \in \mathbb{R}$ , then

$$g^{-1}\{0\} = \{x; f(x) = 0\} \cup \{x; f(x) = \pm\infty\} = f^{-1}\{0\} \cup f^{-1}\{\pm\infty\}.$$

Hence, for  $c > 0$ ,

$$\begin{aligned} g^{-1}\{(0, c)\} &= g^{-1}\{(-\infty, 0)\} \cup g^{-1}\{0\} \cup g^{-1}\{(0, \infty)\} \\ &= f^{-1}\{(-\infty, 0)\} \cup f^{-1}\{0\} \cup f^{-1}\{\pm\infty\} \cup f^{-1}\{(1/c, \infty)\} \\ &= f^{-1}\{(-\infty, 0]\} \cup f^{-1}\{\pm\infty\} \cup f^{-1}\{(1/c, \infty)\}, \end{aligned}$$

which is measurable by the measurability of  $f$  (see Problem 2.1.4). Finally,  $g^{-1}\{\pm\infty\} = \emptyset$ , and it follows by Theorem 2.1.1 that  $g$  is measurable.

## Section 2.3 – Egoroff's Theorem

### Problems

#### 2.3.2

Let  $\{f_n\}$  be a sequence of measurable functions in a finite measure space  $X$ . Suppose that for almost every  $x$ ,  $\{f_n(x)\}$  is a bounded set. Then for any  $\epsilon > 0$  there exist a positive number  $c$  and a measurable set  $E$  with  $\mu(X - E) < \epsilon$ , such that  $|f_n(x)| \leq c$  for all  $x \in E$ ,  $n = 1, 2, \dots$ .

*Solution.* The definition we have for ‘bounded set’ applies to metric spaces, and it does not make much sense here since the functions  $f_n$  may be extended real-valued. Hence we will assume that ‘ $\{f_n(x)\}$  is a bounded set’ means that  $\sup_n |f_n(x)| < \infty$ .

Let  $g = \sup_n |f_n|$ , and note that  $g$  is measurable by Problem 2.1.9 and Theorem 2.2.3. Let  $F = \{x; g(x) < \infty\}$ . Notice that  $g(x) < \infty$  if and only if  $\{f_n(x)\}$  is bounded. Hence  $\mu(X - F) = 0$ .

For  $k = 1, 2, \dots$ , define  $F_k = \{x; g(x) \leq k\}$ . Then  $F_1 \subset F_2 \subset \dots$  and  $\lim_k F_k = \bigcup_{k=1}^{\infty} F_k = F$ . By Theorem 1.2.1(iv),

$$\lim_k \mu(X - F_k) = \mu(X - F) = 0.$$

Given any  $\epsilon > 0$ , there exists a positive integer  $K$  such that  $\mu(X - F_k) < \epsilon$  for all  $k \geq K$ . In particular  $\mu(X - F_K) < \epsilon$ , and  $g(x) \leq K$  for all  $x \in F_K$ , which means that  $|f_n(x)| \leq K$  for all  $x \in F_K$ .

## Section 2.4 – Convergence in Measure

### Problems

#### 2.4.3

Prove the following result (which immediately yields another proof of Corollary 2.4.2): Let  $f_n$  ( $n = 1, 2, \dots$ ) and  $f$  be a.e. real-valued measurable functions in a finite measure space. For any  $\epsilon > 0$ ,  $n \geq 1$ , let

$$E_n(\epsilon) = \{x; |f_n(x) - f(x)| \geq \epsilon\}.$$

Then  $\{f_n\}$  converges a.e. to  $f$  if and only if

$$\lim_{n \rightarrow \infty} \mu \left[ \bigcup_{m=n}^{\infty} E_m(\epsilon) \right] = 0 \quad \text{for any } \epsilon > 0. \quad (2.4.2)$$

[Hint: Let  $F = \{x; \{f_n(x)\} \text{ is not convergent to } f(x)\}$ . Then  $F = \bigcup_{k=1}^{\infty} \overline{\lim_n E_n(1/k)}$ . Show that  $\mu(F) = 0$  if and only if (2.4.2) holds.]

*Solution.* Define

$$F = \bigcup_{k=1}^{\infty} \overline{\lim_n E_n \left( \frac{1}{k} \right)} = \bigcup_{k=1}^{\infty} \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{k} \right).$$

Note that

$$\begin{aligned} x \in F &\iff \exists k, \forall n, \exists m \geq n, |f_m(x) - f(x)| \geq \frac{1}{k} \\ &\iff \neg \left( \forall k, \exists n, \forall m \geq n, |f_m(x) - f(x)| < \frac{1}{k} \right) \\ &\iff f_n(x) \not\rightarrow f(x), \end{aligned}$$

so that

$$F = \{x; f_n(x) \not\rightarrow f(x)\}.$$

Suppose (2.4.2) holds. Fix  $\delta > 0$ . For every positive integer  $k$ , there exists a positive integer  $n_k$  such that  $n \geq n_k$  implies

$$\mu \left[ \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{k} \right) \right] < \frac{\delta}{2^k}.$$

By subadditivity and monotonicity,

$$\begin{aligned}\mu(F) &= \mu \left[ \bigcup_{k=1}^{\infty} \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{k} \right) \right] \leq \sum_{k=1}^{\infty} \mu \left[ \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{k} \right) \right] \\ &\leq \sum_{k=1}^{\infty} \mu \left[ \bigcup_{m=n_k}^{\infty} E_m \left( \frac{1}{k} \right) \right] < \sum_{k=1}^{\infty} \frac{\delta}{2^k} = \delta.\end{aligned}$$

Since  $\delta$  was arbitrary,  $\mu(F) = 0$ , and it follows that  $f_n \rightarrow f$  a.e.

Conversely, suppose  $f_n \rightarrow f$  a.e., so that  $\mu(F) = 0$ . By monotonicity and Theorem 1.2.2,

$$0 = \mu(F) = \mu \left[ \bigcup_{k=1}^{\infty} \overline{\lim}_n E_n \left( \frac{1}{k} \right) \right] \geq \mu \left[ \overline{\lim}_n E_n \left( \frac{1}{l} \right) \right] \geq \overline{\lim}_n \mu \left[ E_n \left( \frac{1}{l} \right) \right]$$

for all positive integers  $l$ . But of course  $\overline{\lim}_n \mu [E_n (1/l)] \geq \underline{\lim}_n \mu [E_n (1/l)] \geq 0$  since  $\mu$  is nonnegative, so  $\lim_n \mu [E_n (1/l)]$  exists and is equal to zero. Note that the sets  $\bigcup_{m=n}^{\infty} E_m (1/l)$  are decreasing, so their limit as  $n \rightarrow \infty$  exists. Hence we can apply Corollary 1.2.3 and monotonicity to find that

$$\begin{aligned}\lim_n \mu \left[ \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{l} \right) \right] &= \mu \left[ \lim_n \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{l} \right) \right] = \mu \left[ \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{l} \right) \right] \\ &\leq \mu \left[ \bigcup_{k=1}^{\infty} \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{k} \right) \right] = \mu(F) = 0.\end{aligned}$$

Finally, given  $\epsilon > 0$ , note that

$$E_n(\epsilon) \subset E_n \left( \frac{1}{\lceil 1/\epsilon \rceil} \right).$$

Hence

$$\lim_n \mu \left[ \bigcup_{m=n}^{\infty} E_m(\epsilon) \right] \leq \lim_n \mu \left[ \bigcup_{m=n}^{\infty} E_m \left( \frac{1}{\lceil 1/\epsilon \rceil} \right) \right] \leq 0$$

by monotonicity, and (2.4.2) follows.

#### 2.4.4

Let  $X$  be the set of all positive integers,  $\mathcal{A}$  the class of all subsets of  $X$ , and  $\mu(E)$  (for any  $E \in \mathcal{A}$ ) the number of points in  $E$ . Prove that in this measure space, convergence in measure is equivalent to uniform convergence.

*Solution.* Uniform convergence always implies convergence in measure. Conversely, suppose  $(f_n)$  converges in measure to  $f$ . Given any  $\epsilon > 0$ , there exists a positive integer  $N$  such that  $n \geq N$  implies

$$\mu [\{x; |f_n(x) - f(x)| \geq \epsilon\}] < 1.$$

That is, for  $n \geq N$  the set  $\{x; |f_n(x) - f(x)| \geq \epsilon\}$  is empty, which in particular means that  $\sup_x |f_n(x) - f(x)| \leq \epsilon$ . It follows that  $f_n \rightarrow f$  uniformly.

## Section 2.5 – Integrals of Simple Functions

### Problems

#### 2.5.2

An integrable simple function  $f$  is equal a.e. to zero if and only if  $\int_E f d\mu = 0$  for any measurable set  $E$ .

*Solution.* Let  $f$  be an integrable simple function. Then  $f$  can be written in the form

$$f = \sum_{i=1}^n \alpha_i \chi_{E_i},$$

for mutually disjoint sets  $E_1, \dots, E_n$ , with all  $\alpha_i \neq 0$ , and all  $\mu(E_i) < \infty$ .

Suppose  $f = 0$  a.e., and let  $E$  be any measurable set. By Theorem 2.5.1(b) and (g),

$$0 \leq \int_E f d\mu \leq \int f d\mu = \sum_{i=1}^n \alpha_i \mu(E_i).$$

But  $\mu(E_i) = 0$  since  $f = 0$  a.e., so  $\int_E f d\mu = 0$ .

Conversely, suppose  $\int_E f d\mu = 0$  for all measurable sets  $E$ . Then

$$\alpha_i \mu(E_i) = \int_{E_i} f d\mu = 0,$$

so that  $\mu(E_i) = 0$ , for all  $i \in \{1, \dots, n\}$ . It follows that  $f = 0$  a.e.

## Section 2.6 – Definition of the Integral

### Problems

#### 2.6.3

Let  $f$  be a measurable function. Prove that  $f$  is integrable if and only if  $f^+$  and  $f^-$  are integrable, or if and only if  $|f|$  is integrable.

*Solution.* Let  $f$  be measurable. We must prove the equivalence of the following statements:

- (i)  $f$  is integrable.
- (ii)  $f^+$  and  $f^-$  are integrable.
- (iii)  $|f|$  is integrable.

We will first show that (iii)  $\implies$  (ii), then that (ii)  $\implies$  (i), and finally that (i)  $\implies$  (iii).

Suppose that  $|f|$  is integrable. Let  $E = \{x; f(x) \geq 0\} = f^{-1}[0, \infty)$ , and note that  $E$  is measurable since  $f$  is. There exists a Cauchy in the mean sequence  $(g_n)$  of integrable simple functions converging to  $|f|$  a.e., and the sequence  $(\chi_E g_n)$

is easily seen to satisfy the corresponding properties with respect to  $f^+$ . Since  $f^+$  is measurable by Problem 2.1.8, this implies that it is integrable. The proof that  $f^-$  is integrable is similar.

Next, suppose that  $f^+$  and  $f^-$  are integrable. Then there exist Cauchy in the mean sequences  $(g_n)$  and  $(h_n)$  of integrable simple functions converging a.e. to  $f^+$  and  $f^-$ , respectively. Define a new sequence  $(f_n)$  of integrable simple functions by  $f_n = g_n - h_n$ . Then  $(f_n)$  is Cauchy in the mean, since

$$|f_n - f_m| = |g_n - h_n - g_m + h_m| \leq |g_n - g_m| + |h_n - h_m|.$$

It also converges to  $f$  a.e. since

$$|f_n - f| = |g_n - h_n - f^+ + f^-| \leq |g_n - f^+| + |h_n - f^-|.$$

It follows that  $f$  is integrable.

Finally, assume that  $f$  is integrable. There is a Cauchy in the mean sequence  $(f_n)$  of integrable simple functions converging to  $f$  a.e. The sequence  $(|f_n|)$  consists of integrable simple functions. It is Cauchy in the mean since

$$||f_n| - |f_m|| \leq |f_n - f_m|,$$

and it converges to  $|f|$  a.e. since

$$||f_n| - |f|| \leq |f_n - f|.$$

Since  $|f|$  is measurable by Problem 2.1.9, it follows that  $|f|$  is integrable.

#### 2.6.4

Let  $X$  be the measure space described in Problem 2.4.4. Then  $f$  is integrable if and only if the series  $\sum_{n=1}^{\infty} |f(n)|$  is convergent. If  $f$  is integrable, then

$$\int f d\mu = \sum_{n=1}^{\infty} f(n).$$

*Solution.* Suppose  $f$  is integrable. Then there is a Cauchy in the mean sequence  $(f_n)$  of integrable simple functions converging to  $f$  a.e. We saw in the previous problem that this implies that  $|f|$  is integrable, and that  $(|f_n|)$  is a Cauchy in the mean sequence of integrable simple functions converging to  $|f|$  a.e. Note that in this particular space convergence a.e. is the same as convergence everywhere (since the only subset with measure zero is  $\emptyset$ ).

By Theorem 2.5.1(h),

$$\int |f_n| d\mu = \sum_{i=1}^{\infty} \int_{\{i\}} |f_n| d\mu = \sum_{i=1}^{\infty} |f_n(i)|.$$

Hence, in particular,

$$\int |f| d\mu = \lim_{n \rightarrow \infty} \sum_{i=1}^{\infty} |f_n(i)|.$$

Given any positive integer  $m$ , there exists  $n'$  such that

$$|f(i) - f_{n'}(i)| < 1/m \quad (i = 1, 2, \dots, m)$$

(since  $f_n \rightarrow f$ ) and

$$\left| \sum_{i=1}^{\infty} |f_{n'}(i)| - \int |f| d\mu \right| < 1$$

(since  $\sum_i |f_n(i)| \rightarrow \int |f| d\mu$ ). Thus

$$\sum_{i=1}^m |f(i)| \leq \sum_{i=1}^m |f(i) - f_{n'}(i)| + \sum_{i=1}^m |f_{n'}(i)| < 1 + \sum_{i=1}^{\infty} |f_{n'}(i)| < 2 + \int |f| d\mu,$$

and it follows that the series  $\sum_{i=1}^{\infty} |f(i)|$  converges (to a finite number).

Conversely, assume that the series  $\sum_{i=1}^{\infty} |f(i)|$  converges. Define a sequence of integrable simple functions  $(g_n)$  by

$$g_n = \sum_{i=1}^n f(i) \chi_{\{i\}}.$$

It is clear that  $g_n \rightarrow f$  everywhere. Moreover, if  $m > n$ , then

$$\int |g_m - g_n| d\mu = \int \left| \sum_{i=n+1}^m f(i) \chi_{\{i\}} \right| d\mu = \sum_{i=n+1}^m |f(i)| \leq \sum_{i=n+1}^{\infty} |f(i)|.$$

The right-hand side goes to zero as  $n \rightarrow \infty$  since  $\sum_{i=1}^{\infty} |f(i)|$  is convergent, which means that  $\int |g_m - g_n| d\mu \rightarrow 0$  as  $n, m \rightarrow \infty$ ; i.e.,  $(g_n)$  is Cauchy in the mean. It follows that  $f$  is integrable, with

$$\int f d\mu = \lim_{n \rightarrow \infty} \int g_n d\mu = \lim_{n \rightarrow \infty} \sum_{i=1}^n f(i) = \sum_{i=1}^{\infty} f(i).$$

## Section 2.7 – Elementary Properties of Integrals

### Problems

#### 2.7.3

Let  $f$  be an integrable function. Prove: (a) if  $\int_E f d\mu \geq 0$  for all measurable sets  $E$ , then  $f \geq 0$  a.e.; (b) if  $\mu(X) < \infty$  and if  $\int_E f d\mu \leq \mu(E)$  for all measurable sets  $E$ , then  $f \leq 1$  a.e.

*Solution.*

(a) Let  $F = \{x; f(x) < 0\}$ . Then  $-f$  is positive on  $F$ , so

$$-\int_F f d\mu = \int_F (-f) d\mu \geq 0$$

by Theorem 2.7.1(b). But

$$\int_F f d\mu \geq 0$$

by our hypothesis on  $f$ , so

$$\int_F f d\mu = 0,$$

and it follows by Theorem 2.7.5 that  $\mu(F) = 0$ . That is,  $f \geq 0$  a.e.

(b) Let  $G = \{x : f(x) > 1\}$ . Then  $f - 1$  is positive on  $G$ , so

$$\int_G (f - 1) d\mu \geq 0$$

by Theorem 2.7.1(b). Note that  $\chi_G$  is integrable since  $\mu(G) \leq \mu(X) < \infty$ . Hence we also have

$$\int_G (f - 1) d\mu = \int_G f d\mu - \int_G d\mu = \int_G f d\mu - \mu(G) \leq 0,$$

with the inequality following from our hypothesis on  $f$ . Thus

$$\int_G (f - 1) d\mu = 0,$$

and Theorem 2.7.5 yields  $\mu(G) = 0$ . That is,  $f \leq 1$  a.e.

## Section 2.8 – Sequences of Integrable Functions

### Problems

#### 2.8.1

A measurable function  $f$  is called a *null function* if  $f = 0$  a.e. We shall say that  $f$  is *equivalent* to  $g$  (and write  $f \sim g$ ) if  $f - g$  is a null function. Denote by  $\bar{f}$  the class of all measurable functions that are equivalent to  $f$ . We denote by  $L^1(X, \mathcal{A}, \mu)$ , or, more briefly, by  $L^1(X, \mu)$ , the set of all classes  $\bar{f}$  for which  $f$  is integrable, and define on it the function

$$\rho(\bar{f}, \bar{g}) = \rho(f, g) = \int |f - g| d\mu.$$

[Note that if  $f_0 \in \bar{f}$ ,  $g_0 \in \bar{g}$ , then  $\rho(f_0, g_0) = \rho(f, g)$ .] Prove that  $L^1(X, \mu)$  is a complete metric space with the metric  $\rho$ .

*Solution.* Let  $(\bar{f}_n)$  be a Cauchy sequence in  $L^1(X, \mu)$ . Then

$$\int |f_m - f_n| d\mu = \rho(\bar{f}_m, \bar{f}_n) \rightarrow 0$$

as  $m, n \rightarrow \infty$ , so the sequence  $(f_n)$  (of representative functions) is Cauchy in the mean. By Theorem 2.8.3 there is an integrable function  $f$  such that  $f_n \rightarrow f$  in the mean. Hence

$$\rho(\bar{f}_n, \bar{f}) = \int |f_n - f| d\mu \rightarrow 0$$

as  $n \rightarrow \infty$ . That is,  $\bar{f}_n \rightarrow \bar{f}$  in  $L^1(X, \mu)$ .

## 2.8.2

TODO.



# Chapter 3 – Metric Spaces

## Section 3.1 – Topological and Metric Spaces

### Problems

#### 3.1.1

Prove that if  $(X, \rho)$  is a metric space, and if

$$\hat{\rho}(x, y) = \frac{\rho(x, y)}{1 + \rho(x, y)},$$

then also  $(X, \hat{\rho})$  is a metric space. [*Hint:* Cf. the proof of (3.1.3).]

*Solution.* The only nonobvious property is the triangle inequality. Let  $x, y, z$  be arbitrary points of  $X$ . Since  $t \mapsto t/(1+t)$  is monotone increasing on  $[0, \infty)$ , and since  $\rho(x, z) \leq \rho(x, y) + \rho(y, z)$ , we have

$$\frac{\rho(x, z)}{1 + \rho(x, z)} \leq \frac{\rho(x, y) + \rho(y, z)}{1 + \rho(x, y) + \rho(y, z)}.$$

Moreover, by equation (3.1.3),

$$\frac{\rho(x, y) + \rho(y, z)}{1 + \rho(x, y) + \rho(y, z)} \leq \frac{\rho(x, y)}{1 + \rho(x, y)} + \frac{\rho(y, z)}{1 + \rho(y, z)}.$$

It follows that  $\hat{\rho}(x, z) \leq \hat{\rho}(x, y) + \rho(y, z)$ .

#### 3.1.2

Let  $X, \rho, \hat{\rho}$  be as in Problem 3.1.1. Prove that  $\rho(x_n, x) \rightarrow 0$  if and only if  $\hat{\rho}(x_n, x) \rightarrow 0$ . Give an example showing that  $\rho$  and  $\hat{\rho}$  are not equivalent in general.

*Solution.* It is clear that  $\rho(x_n, x) \rightarrow 0$  implies  $\hat{\rho}(x_n, x) \rightarrow 0$ , since  $\hat{\rho} \leq \rho$ .

Conversely, suppose  $\hat{\rho}(x_n, x) \rightarrow 0$ . Given any  $\epsilon > 0$ , there is a positive integer  $N$  such that

$$\hat{\rho}(x_n, x) < \frac{\epsilon}{1 + \epsilon} \quad (n \geq N).$$

Substituting the definition of  $\hat{\rho}$  and rearranging yields  $\rho(x_n, x) < \epsilon$ .

If  $\rho$  and  $\hat{\rho}$  are equivalent then, in particular, there exists a positive constant  $\beta$  such that

$$\frac{\rho(x, y)}{\hat{\rho}(x, y)} \leq \beta$$

whenever  $x \neq y$ . But

$$\frac{\rho(x, y)}{\hat{\rho}(x, y)} = 1 + \rho(x, y),$$

so this is impossible if  $X$  is unbounded (w.r.t.  $\rho$ ), say if  $X = \mathbb{R}^n$  and  $\rho$  is the Euclidean metric.

### 3.1.6

Prove that the spaces  $l^1, s, c, c_0$  are separable metric spaces.

*Solution.* For each of the spaces  $X$  we will take an arbitrary element  $x \in X$  and demonstrate that every  $\epsilon$ -ball around  $x$  contains a point  $y$  of a certain countable subset  $Y \subset X$ . It will then follow that  $Y$  is dense in  $X$ , and hence that  $X$  is separable.

Let  $x = (x_i) \in l^1$ . Fix  $\epsilon > 0$ . Since  $\sum_i |x_i| < \infty$ , there exists  $n$  such that

$$\sum_{i=n+1}^{\infty} |x_i| < \frac{\epsilon}{2}.$$

For  $i = 1, \dots, n$ , choose  $y_i \in \mathbb{Q}$  (or  $y_i$  with rational real and imaginary parts in the complex case) such that  $|x_i - y_i| < \epsilon/2n$ , and let  $y = (y_1, \dots, y_n, 0, 0, \dots)$ . Then

$$\rho(x, y) = \sum_{i=1}^n |x_i - y_i| + \sum_{i=n+1}^{\infty} |x_i| < \epsilon.$$

Moreover,  $y$  is an element of the subset of  $l^1$  consisting of sequences with rational components, with only finitely many being nonzero. This subset is easily seen to be countable, and it follows that  $l^1$  is separable.

Next, let  $x = (x_i) \in s$ , and fix  $\epsilon > 0$ . Choose  $n$  such that

$$\sum_{i=n+1}^{\infty} \frac{1}{2^i} \frac{|x_i|}{1 + |x_i|} < \frac{\epsilon}{2}.$$

For  $i = 1, \dots, n$ , choose  $y_i \in \mathbb{Q}$  such that  $|x_i - y_i| < \epsilon/2$ , and let  $y = (y_1, \dots, y_n, 0, 0, \dots)$ . Then

$$\rho(x, y) = \sum_{i=1}^n \frac{1}{2^i} \frac{|x_i - y_i|}{1 + |x_i - y_i|} + \sum_{i=n+1}^{\infty} \frac{1}{2^i} \frac{|x_i|}{1 + |x_i|} < \epsilon.$$

Similarly to the above, it follows that  $s$  is dense.

Finally, let  $x = (x_i) \in c$ . (This argument will also cover  $c_0$ .) Let  $\xi = \lim_i x_i$  and fix  $\epsilon > 0$ . Choose  $n$  such that  $|x_i - \xi| < \epsilon/2$  for all  $i \geq n$ . For  $i = 1, \dots, n-1$ ,

choose  $y_i \in \mathbb{Q}$  such that  $|x_i - y_i| < \epsilon$ . Also choose  $\eta \in \mathbb{Q}$  such that  $|\xi - \eta| < \epsilon/2$  (take  $\eta = \xi = 0$  in the  $c_0$  case), so that

$$|x_i - \eta| \leq |x_i - \xi| + |\xi - \eta| < \epsilon$$

for all  $i \geq n$ . Let  $y = (y_1, \dots, y_{n-1}, \eta, \eta, \dots)$ . Then

$$\rho(x, y) = \sup_i |x_i - y_i| \leq \epsilon.$$

It follows that  $c$  is separable (and  $c_0$  as well).

### 3.1.10

If  $\rho(x_n, x) \rightarrow 0$ ,  $\rho(y_n, y) \rightarrow 0$ , then  $\rho(x_n, y_n) \rightarrow \rho(x, y)$ .

*Solution.* The triangle inequality yields

$$\rho(x_n, y_n) \leq \rho(x_n, x) + \rho(x, y) + \rho(y, y_n)$$

and

$$\rho(x, y) \leq \rho(x, x_n) + \rho(x_n, y_n) + \rho(y_n, y).$$

Hence

$$|\rho(x_n, y_n) - \rho(x, y)| \leq \rho(x_n, x) + \rho(y_n, y),$$

and the result follows.

## Section 3.2 – $L^p$ Spaces

### Problems

#### 3.2.4

Prove that  $l^p$  is separable if  $1 \leq p < \infty$ .

*Solution.* The argument is of the same sort as in Problem 3.1.6. Given  $x = (x_i) \in l^p$  ( $1 \leq p < \infty$ ) and  $\epsilon > 0$ , choose  $n$  such that

$$\sum_{i=n+1}^{\infty} |x_i|^p < \frac{\epsilon^p}{2}.$$

(This is possible since  $\sum_i |x_i|^p < \infty$  for  $x \in l^p$ .) For  $i = 1, \dots, n$ , choose  $y_i \in \mathbb{Q}$  such that  $|x_i - y_i|^p < \epsilon^p/2n$  and let  $y = (y_1, \dots, y_n, 0, 0, \dots)$ . Then  $\|x - y\|_p < \epsilon$ , and the conclusion follows as in Problem 3.1.6.

### 3.2.5

Prove that  $l^p$  is not a metric space if  $0 < p < 1$ .

*Solution.* Let  $x = (1, 0, 0, \dots)$ ,  $y = (0, 0, \dots)$ ,  $z = (0, 1, 0, 0, \dots)$ . Then

$$\|x - z\|_p = 2^{1/p}$$

while

$$\|x - y\|_p + \|y - z\|_p = 2.$$

But  $2^{1/p} > 2$  if  $0 < p < 1$ , so the triangle inequality does not hold.

### 3.2.6

Prove that the space  $C[a, b]$  with the metric

$$\rho(f, g) = \int_a^b |f(t) - g(t)| dt$$

is not a complete metric space.

*Solution.* Note that the integral defining  $\rho$  is guaranteed to exist by Theorem 2.11.1 and the extreme value theorem from elementary analysis, and it can be taken in the Riemann sense. It is easily seen that  $\rho$  is indeed a metric.

For simplicity, let us assume  $a = 0$  and  $b = 1$ . (The general argument is identical modulo a coordinate transformation.) Define a sequence of functions in  $C[0, 1]$  by  $f_n(x) = x^n$ . Clearly  $(f_n)$  converges pointwise to the discontinuous function

$$f(x) = \begin{cases} 0 & (0 \leq x < 1), \\ 1 & (x = 1). \end{cases}$$

For  $n \geq m$  we have  $f_n \geq f_m$ , hence

$$\rho(f_n, f_m) = \int_0^1 f_n(t) dt - \int_0^1 f_m(t) dt = \frac{1}{n+1} - \frac{1}{m+1},$$

which goes to zero as  $n, m \rightarrow \infty$ . Thus  $(f_n)$  is a Cauchy sequence in  $C[0, 1]$  which converges to a function  $f$  which is not in  $C[0, 1]$ , so  $(C[0, 1], \rho)$  is not a complete metric space.

## Section 3.4 – Complete Metric Spaces

### Problems

#### 3.4.5

A set  $Y$  in a metric space  $X$  is said to be *of the first category in  $X$*  if it is contained in a countable union of nowhere dense sets of  $X$ . If  $Y$  is not of the first category in  $X$ , then it is said to be *of the second category in  $X$* . The real line with the Euclidean metric is a space of the second category. Prove, however, that, as a subset of the Euclidean plane, the real line is a set of the first category.

*Solution.* We identify the real line with the subset  $L = \{(x, y) \in \mathbb{R}^2 : y = 0\}$ , with the induced metric topology. Indeed  $L$  is homeomorphic to  $\mathbb{R}$  via the map  $(x, 0) \mapsto x$ , as can easily be shown.

The sets of the form  $B((0, 0), r) \cap L = \{(x, 0) : |x| < r\}$  are easily seen to be nowhere dense in  $\mathbb{R}^2$ , and we have

$$L = \bigcup_{n=1}^{\infty} B((0, 0), n) \cap L.$$

Hence  $L$  is of the first category in  $\mathbb{R}^2$ .

### 3.4.7

Let  $f(x)$  be a real-valued function on the real line. Prove that there is a nonempty interval  $(a, b)$  and a positive number  $c$  such that for any  $x \in (a, b)$  there is a sequence  $\{x_n\}$  such that  $x_n \rightarrow x$  and  $|f(x_n)| \leq c$ .

*Solution.* Note that

$$\mathbb{R} = |f|^{-1}(\mathbb{R}) = |f|^{-1}\left(\bigcup_{n=1}^{\infty} (-\infty, n]\right) = \bigcup_{n=1}^{\infty} |f|^{-1}((-\infty, n]).$$

Since  $\mathbb{R}$  is of the second category (by Theorem 3.4.2), the sets  $|f|^{-1}((-\infty, n])$  cannot all be nowhere dense. Hence there is a positive integer  $c$  such that the closure of  $|f|^{-1}((-\infty, c])$  has nonempty interior. On account of being a nonempty open set, this interior contains an open interval  $(a, b)$ .

Let  $x \in (a, b)$ . Then, since  $x$  is contained in the closure of  $|f|^{-1}((-\infty, c])$ , every neighborhood of  $x$  contains a point of  $|f|^{-1}((-\infty, c])$ . It follows that we can construct a sequence  $(x_n)$  in  $|f|^{-1}((-\infty, c])$  converging to  $x$ , and this sequence will satisfy  $|f(x_n)| \leq c$  for all  $n$ .

## Section 3.5 – Compact Metric Spaces

### Problems

#### 3.5.4

A subset  $F$  of a compact metric space is compact if and only if it is closed.

*Solution.* Let  $X$  be a compact metric space, and  $F$  a subset of  $X$ . If  $F$  is compact then it is also closed by Corollary 3.5.5.

Conversely, suppose that  $F$  is closed. Let  $\mathcal{C}$  be any open cover of  $F$ . The collection  $\mathcal{C} \cup \{X - F\}$  is an open cover of  $X$ , hence it has a finite subcover, consisting of some sets  $E_1, \dots, E_n \in \mathcal{C}$  and perhaps  $\{X - F\}$ . The sets  $E_1, \dots, E_n$  cover  $F$ , hence  $F$  is compact.

### 3.5.5

A subset  $Y$  of a metric space is totally bounded if and only if its closure  $\overline{Y}$  is totally bounded.

*Solution.* Clearly  $Y$  is totally bounded whenever  $\overline{Y}$  is. Conversely, suppose  $Y$  is totally bounded, and let  $\epsilon > 0$  be given. Then  $Y$  admits a finite  $\epsilon/2$ -covering,

$$\{B(x_1, \epsilon/2), \dots, B(x_n, \epsilon/2)\}.$$

The union of the closed balls,

$$\bigcup_{i=1}^n \overline{B}(x_i, \epsilon/2),$$

is a closed set containing  $Y$ , hence it contains  $\overline{Y}$ . It follows that

$$\overline{Y} \subset \bigcup_{i=1}^n \overline{B}(x_i, \epsilon/2) \subset \bigcup_{i=1}^n B(x_i, \epsilon),$$

so  $\overline{Y}$  admits a finite  $\epsilon$ -covering.

### 3.5.6

The intersection of any number of compact subsets of a metric space is a compact space.

*Solution.* Let  $\mathcal{C}$  be a collection of compact subsets of a metric space  $X$ . By Corollary 3.5.5 each  $C \in \mathcal{C}$  is closed, hence the intersection  $K = \bigcap \mathcal{C}$  is closed as well. Let  $C'$  be some particular member of  $\mathcal{C}$ . Then  $K$  is a closed subset of the compact space  $C'$ , thus itself compact by Problem 3.5.4.

Remark: Really the above shows that  $K$  is compact *in*  $C'$ . However, one easily proves the general result that if  $F$  is compact in  $G$ , and  $G$  is compact in  $H$ , then  $F$  is compact in  $H$ . Hence  $K$  is indeed compact in  $X$ .

### 3.5.8

Show that a metric space is compact if and only if it has the following property: for every collection of closed subsets  $\{F_\alpha\}$ , if any finite subcollection has nonempty intersection, then the whole collection has a nonempty intersection.

*Solution.* The wording of the problem is potentially misleading. Replace “any” with “every” to make it nonambiguous. Also, let us introduce some useful terminology. A collection of subsets of a topological space is said to have the *finite intersection property* iff every finite subcollection has nonempty intersection. Our task is therefore to show that a metric space  $X$  is compact iff every collection  $\mathcal{F}$  of closed subsets of  $X$  having the finite intersection property has nonempty intersection. However, the proof we will use is valid for general topological spaces, not just metric spaces.

Let  $X$  be a topological space, and suppose first that  $X$  is compact. Let  $\mathcal{F}$  be a collection of closed subsets of  $X$  with the finite intersection property. Assume towards a contradiction that  $\bigcap \mathcal{F} = \emptyset$ . Then

$$\bigcup_{F \in \mathcal{F}} (X - F) = X - \bigcap_{F \in \mathcal{F}} F = X,$$

so  $\{X - F; F \in \mathcal{F}\}$  is an open cover of  $X$ . Since  $X$  is compact, there are sets  $F_1, \dots, F_n \in \mathcal{F}$  such that

$$X = \bigcup_{i=1}^n (X - F_i) = X - \bigcap_{i=1}^n F_i.$$

But then  $\bigcap_{i=1}^n F_i = \emptyset$ , contradicting the hypothesis that  $\mathcal{F}$  has the finite intersection property. We conclude that  $\bigcap \mathcal{F} \neq \emptyset$ .

Conversely, suppose that  $X$  has the property described in the problem statement: every collection of closed subsets of  $X$  with the finite intersection property has nonempty intersection. Let  $\mathcal{U}$  be any open cover of  $X$ , and assume towards a contradiction that  $\mathcal{U}$  has no finite subcover. Then

$$\bigcap_{i=1}^n (X - U_i) = X - \bigcup_{i=1}^n U_i \neq \emptyset$$

for every finite subcollection  $\{U_1, \dots, U_n\} \subset \mathcal{U}$ . Hence  $\mathcal{F} = \{X - U; U \in \mathcal{U}\}$  is a collection of closed subsets with the finite intersection property. By hypothesis  $\mathcal{F}$  has nonempty intersection, so

$$\emptyset = X - \bigcup_{U \in \mathcal{U}} U = \bigcap_{U \in \mathcal{U}} (X - U) = \bigcap_{F \in \mathcal{F}} F \neq \emptyset,$$

a contradiction. We conclude that  $\mathcal{U}$  does indeed have a finite subcover, and that  $X$  is compact.

# Chapter 4 – Elements of Functional Analysis in Banach Spaces

## Section 4.1 – Linear Normed Spaces

### Problems

#### 4.1.4

If  $\{x_n\}$  is a convergent sequence in a normed linear space, with limit  $x$ , then also the sequence with elements  $(x_1 + \cdots + x_n)/n$  is convergent to  $x$ .

*Solution.* For  $n = 1, 2, \dots$ , define

$$\sigma_n = \frac{x_1 + \cdots + x_n}{n} = \frac{1}{n} \sum_{i=1}^n x_i.$$

Let  $\epsilon > 0$  be given. Since  $(x_n)$  is convergent there exists a positive integer  $m$  such that  $\|x_n - x\| < \epsilon/2$  for all  $n \geq m + 1$ . For such  $n$  we have

$$\begin{aligned} \|\sigma_n - x\| &= \left\| \frac{1}{n} \sum_{i=1}^n (x_i - x) \right\| \\ &\leq \frac{1}{n} \left\| \sum_{i=1}^m (x_i - x) \right\| + \frac{1}{n} \left\| \sum_{i=m+1}^n (x_i - x) \right\| \\ &\leq \frac{1}{n} \left\| \sum_{i=1}^m (x_i - x) \right\| + \frac{1}{n} \sum_{i=m+1}^n \|x_i - x\| \\ &\leq \frac{1}{n} \left\| \sum_{i=1}^m (x_i - x) \right\| + \frac{n - m - 1}{n} \cdot \frac{\epsilon}{2}. \end{aligned}$$

The first term on the right-hand side goes to zero as  $n \rightarrow \infty$ , and the other goes to  $\epsilon/2$ . Hence  $\|\sigma_n - x\| < \epsilon$  for sufficiently large  $n$ , and it follows that  $\sigma_n \rightarrow x$ .



#### 4.1.6

A normed linear space is a Banach space if the following property is satisfied: every absolutely convergent series is convergent.

*Solution.* Let  $X$  be a normed linear space with the property that every absolutely convergent series is convergent. Let  $(x_n)$  be a Cauchy sequence in  $X$ , and choose integers  $N_1 < N_2 < \dots$  such that  $m, n \geq N_k$  implies  $\|x_m - x_n\| < 2^{-k}$ . Define a new sequence  $(y_k)$  by  $y_1 = x_{N_1}$  and  $y_k = x_{N_k} - x_{N_{k-1}}$  for  $k > 1$ . Then

$$\sum_{k=1}^{\infty} \|y_k\| = \|x_{N_1}\| + \sum_{k=2}^{\infty} \|x_{N_k} - x_{N_{k-1}}\| \leq \|x_{N_1}\| + \sum_{k=1}^{\infty} 2^{-k} = \|x_{N_1}\| + 1.$$

It follows by our hypothesis on  $X$  that

$$x_{N_k} = \sum_{i=1}^k y_i$$

converges to some  $x \in X$ . Given  $\epsilon > 0$ , choose an integer  $j \geq 1$  such that  $2^{-j} < \epsilon$ , and  $k > j$  such that

$$\|x_{N_k} - x\| < \epsilon - 2^{-j}.$$

Then, for all  $n \geq N_k$ ,

$$\|x_n - x\| \leq \|x_n - x_{N_k}\| + \|x_{N_k} - x\| < 2^{-k} + \epsilon - 2^{-j} < \epsilon.$$

We conclude that every Cauchy sequence in  $X$  is convergent, and hence that  $X$  is a Banach space.

## Section 4.2 – Subspaces and Bases

### Problems

#### 4.2.6

If a linear vector space is infinite-dimensional, then there exist on it norms that are not equivalent. [*Hint:* Let  $\{y_\alpha\}$  be a Hamel basis and define norms by  $\|x\|^2 = \sum_{\alpha} c_{\alpha} |\lambda_{\alpha}|^2$ , where  $x$  has the form (4.2.2) and  $c_{\alpha}$  are positive numbers.]

*Solution.* Let  $\{y_{\alpha}\}_{\alpha \in I}$  be a Hamel basis for an infinite-dimensional linear space  $X$ . For every  $x = \sum_{\alpha \in I} \lambda_{\alpha} y_{\alpha}$ , define

$$\|x\|_1 = \left( \sum_{\alpha \in I} |\lambda_{\alpha}|^2 \right)^{1/2}.$$

Then  $\|\cdot\|_1$  is easily seen to be a norm on  $X$ .

Let  $\{\alpha_i\}_{i=1}^\infty$  be a countable subset of the index set  $I$ . Define a set  $\{c_\alpha\}_{\alpha \in I}$  of positive constants by  $c_{\alpha_i} = 2^{-i}$  for  $i = 1, 2, \dots$ , and  $c_\alpha = 1$  for  $\alpha \notin \{\alpha_i\}$ . Define another norm  $\|\cdot\|_2$  by

$$\|x\|_2 = \left( \sum_{\alpha \in I} c_\alpha |\lambda_\alpha|^2 \right)^{1/2},$$

for  $x = \sum_{\alpha \in I} \lambda_\alpha y_\alpha$ .

To see that these two norms are not equivalent, note that

$$\|y_{\alpha_i}\|_2^2 = c_{\alpha_i} = 2^{-i} = 2^{-i} \|y_{\alpha_i}\|_1^2$$

for all  $i$ . Hence there exists no  $\beta > 0$  such that  $\|x\|_1 \leq \beta \|x\|_2$  for all  $x \in X$ .

## Section 4.3 – Finite-Dimensional Normed Linear Spaces

### Problems

#### 4.3.1

Let  $X$  be a finite-dimensional linear space. Then any two norms on  $X$  are equivalent. (According to Problem 4.2.6, the assertion is false if  $X$  is infinite-dimensional.)

*Solution.* Let  $e_1, \dots, e_n$  be a basis of  $X$ . For every  $x = \sum_{i=1}^n \lambda_i e_i \in X$ , let

$$\|x\|_1 = \sum_{i=1}^n |\lambda_i|.$$

It is easily verified that  $\|\cdot\|_1$  is a norm on  $X$ . We will show that every norm on  $X$  is equivalent to  $\|\cdot\|_1$ , and hence (by transitivity) that any two norms on  $X$  are equivalent.

Given an arbitrary norm  $\|\cdot\|$ , we must prove the existence of positive constants  $\alpha$  and  $\beta$  such that

$$\alpha \|x\|_1 \leq \|x\| \leq \beta \|x\|_1$$

for all  $x \in X$ . These inequalities hold trivially for  $x = 0$ , so it suffices to consider nonzero  $x$ . In fact it is sufficient to consider  $x$  in the “ $\|\cdot\|_1$ -sphere”  $S_1 = \{x \in X : \|x\|_1 = 1\}$ , where the inequalities reduce to

$$\alpha \leq \|x\| \leq \beta.$$

The inequality for general, nonzero  $x$  then follows upon division by  $\|x\|_1$ .

We start by showing that the map  $x \mapsto \|x\|$  is continuous with respect to the metric  $\rho_1(x, y) = \|x - y\|_1$ . Let  $\epsilon > 0$  be given, and write  $M = \max(\|e_1\|, \dots, \|e_n\|)$ . Given

$$x = \sum_{i=1}^n \lambda_i e_i, \quad y = \sum_{i=1}^n \mu_i e_i$$

satisfying  $\rho_1(x, y) < \epsilon/M$ , we have

$$\|x - y\| \leq \sum_{i=1}^n |\lambda_i - \mu_i| \|e_i\| \leq \sum_{i=1}^n |\lambda_i - \mu_i| M = \rho_1(x, y) M < \epsilon,$$

and continuity follows.

The sphere  $S_1$  is obviously closed and bounded under  $\|\cdot\|_1$ , hence compact by Theorem 4.3.3. By Theorem 3.6.2 and the continuity established above, the map  $x \mapsto \|x\|$  attains a maximum and a minimum on  $S_1$ . Let

$$\alpha = \inf_{x \in S_1} \|x\|, \quad \beta = \sup_{x \in S_1} \|x\|,$$

and note that  $\alpha > 0$  since  $\|x\| = \alpha$  is attained for some nonzero  $x$ . It follows that  $\alpha$  and  $\beta$  are positive constants such that  $\alpha \leq \|x\| \leq \beta$  for all  $x \in S_1$ , so we are done.

#### 4.3.2

Let  $Y$  be a finite-dimensional linear subspace of a normed linear space  $X$ , and let  $x_0 \in X$ ,  $x_0 \notin Y$ . Then there exists a point  $y_0 \in Y$  such that

$$\inf_{y \in Y} \|x_0 - y\| = \|x_0 - y_0\|.$$

*Solution.* Note that  $Y$  is closed by Theorem 4.3.2. Let  $L = \inf_{y \in Y} \|x_0 - y\|$ . For  $n = 1, 2, \dots$ , choose  $y_n \in Y$  such that

$$\|x_0 - y_n\| < L + \frac{1}{n}.$$

Note that  $y_n \in B(x_0, L + 1) \cap Y$  for all  $n$ . This is a bounded subset of  $Y$ , so by Theorems 4.3.3 and 3.5.4, the sequence  $(y_n)$  has a subsequence  $(y_{n_k})$  converging to a point

$$y_0 \in \overline{B(x_0, L + 1) \cap Y} \subset \overline{Y} = Y.$$

For all  $k$  we have

$$\|x_0 - y_0\| \leq \|x_0 - y_{n_k}\| + \|y_{n_k} - y_0\|.$$

The left-hand side of this inequality is bounded below by  $L$ , and the right-hand side converges to  $L$  as  $k \rightarrow \infty$ . It follows that  $\|x_0 - y_0\| = L$ .

### 4.3.3

A norm  $\| \cdot \|$  is said to be *strictly convex* if  $\|x\| = 1$ ,  $\|y\| = 1$ ,  $\|x + y\| = 2$  imply that  $x = y$ . Prove that if the norm of  $X$  is strictly convex, then the point  $y_0$  occurring in the assertion of Problem 4.3.2 is unique.

*Solution.* Let  $\| \cdot \|$  be a strictly convex norm on a linear space  $X$ , and let  $Y$  be a finite-dimensional linear subspace of  $X$ . Let  $x_0 \in X - Y$ ,  $L = \inf_{y \in Y} \|x_0 - y\|$ , and suppose that there are elements  $y_0, y'_0 \in Y$  such that

$$\|x_0 - y_0\| = \|x_0 - y'_0\| = L.$$

It is clear that  $\|(x_0 - y_0)/L\| = \|(x_0 - y'_0)/L\| = 1$ . Moreover,

$$\|(x_0 - y_0)/L + (x_0 - y'_0)/L\| = \frac{2}{L}\|x_0 - (y_0 + y'_0)/2\| \geq \frac{2}{L}L = 2,$$

and

$$\|(x_0 - y_0)/L + (x_0 - y'_0)/L\| \leq \frac{1}{L}(\|x_0 - y_0\| + \|x_0 - y'_0\|) = \frac{1}{L}2L = 2,$$

so  $\|(x_0 - y_0)/L + (x_0 - y'_0)/L\| = 2$ . By strict convexity,

$$\frac{1}{L}(x_0 - y_0) = \frac{1}{L}(x_0 - y'_0),$$

and it follows that  $y_0 = y'_0$ .

### 4.3.4

Prove that the norm of  $L^p(X, \mu)$  is strictly convex if  $1 < p < \infty$ , and is not strictly convex if  $p = 1$  or if  $p = \infty$ .

*Solution.* Let  $f, g \in \mathcal{L}^p(X, \mu)$ ,  $1 < p < \infty$ , and suppose  $\|f\|_p = \|g\|_p = 1$  and  $\|f + g\|_p = 2$ . We then have equality in Minkowski's inequality:

$$\|f + g\|_p = \|f\|_p + \|g\|_p.$$

By Problem 3.2.7 this implies that  $f = 0$  a.e., or  $g = 0$  a.e., or  $f = \lambda g$  a.e. for some positive constant  $\lambda$ . The first two possibilities are ruled out since  $\|f\|_p = \|g\|_p = 1$ , so the third alternative must hold. But then

$$1 = \|f\|_p = |\lambda| \|g\|_p = |\lambda|,$$

so  $\lambda = 1$ . Thus  $f = g$  a.e., so that  $\tilde{f} = \tilde{g}$  in  $L^p(X, \mu)$ . It follows that  $\| \cdot \|_p$  is strictly convex if  $1 < p < \infty$ .

For  $p = 1, \infty$ , surely the problem is supposed to say 'not *necessarily* strictly convex', because we can come up with examples where  $\| \cdot \|_p$  is strictly convex, such as an empty measure space  $X = \emptyset$ , or any space with identically zero measure  $\mu = 0$ . (Perhaps less trivial examples exist.) Hence we will only

demonstrate that there are examples where  $\|\cdot\|_p$  ( $p \in \{1, \infty\}$ ) is not strictly convex.

For  $p = 1$ , consider the  $L^1$ -space  $l^1$ . The sequences  $x = (1, 0, 0, \dots)$  and  $y = (0, 1, 0, 0, \dots)$  satisfy  $\|x\|_1 = \|y\|_1 = 1$  and  $\|x + y\|_1 = 2$ , yet  $x \neq y$ . For  $p = \infty$ , consider the  $L^\infty$ -space  $l^\infty$ . The sequences  $x = (1, 0, 0, \dots)$  and  $y = (1, 1, 1, \dots)$  satisfy  $\|x\|_\infty = \|y\|_\infty = 1$  and  $\|x + y\|_\infty = 2$ , but  $x \neq y$ .

### 4.3.5

Prove that in  $C[a, b]$  the uniform norm is not equivalent to the  $L^p$  norm (for  $1 \leq p < \infty$ ).

*Solution.* For simplicity, let  $a = 0$  and  $b = 1$ . For  $n = 1, 2, \dots$ , define  $f_n : [0, 1] \rightarrow \mathbb{R}$  (or  $\mathbb{C}$ ) by  $f_n(x) = x^n$ . Let  $\|\cdot\|_u$  denote the uniform norm. For all  $n$  we have

$$\|f_n\|_u = \max_{0 \leq x \leq 1} |f_n(x)| = 1$$

and

$$\|f_n\|_p = \left( \int_0^1 x^n dx \right)^{1/p} = (n+1)^{-1/p}.$$

(We are assuming that the  $L^p$  norm is defined with the standard Lebesgue measure.) In particular  $\|f_n\|_p \rightarrow 0$  as  $n \rightarrow \infty$ , whatever the value of  $p$  ( $\neq \infty$ ). It follows that there exists no  $\beta > 0$  such that  $\|f\|_u \leq \beta \|f\|_p$  for all  $f \in C[0, 1]$ , and hence that the two norms are not equivalent.

### 4.3.7

Let  $n$  be a positive integer,  $1 \leq p < \infty$ , and let  $f(x)$  be a continuous function on  $0 \leq x \leq 1$ . Then there exists a unique polynomial  $Q_n$  of degree  $n$  such that for any other polynomial  $P_n$  of degree  $n$

$$\int_0^1 |f(x) - P_n(x)|^p dx > \int_0^1 |f(x) - Q_n(x)|^p dx.$$

*Solution.* There is an error in the problem statement; it should be polynomials of degree  $\leq n$ , not *exactly*  $n$ . To see that the written claim is false, consider the case  $f(x) = 0$  and  $n = 1$ . Then  $f$  can be approximated arbitrarily closely by degree 1 polynomials  $ax + b$ ,  $a \neq 0$ , but no such polynomial will make  $\|f - P\|_p$  vanish completely.

Let  $\mathcal{P}_n$  denote the set of polynomial functions on  $[0, 1]$  with degree  $\leq n$ . Then  $\mathcal{P}_n$  is a linear subspace of the normed linear space  $(C[0, 1], \|\cdot\|_p)$ , and is finite-dimensional since it is spanned by the polynomials  $1, x, x^2, \dots, x^n$ .

If  $f \in \mathcal{P}_n$ , then  $Q = f$  satisfies the claim, since  $\|f - P\|_p > 0$  for all  $P \neq f$  (else  $\|\cdot\|_p$  would not be a norm). If  $f \notin \mathcal{P}_n$ , then the conclusion of Problem 4.3.2 tells us that there exists  $Q \in \mathcal{P}_n$  such that

$$\inf_{P \in \mathcal{P}_n} \|f - P\| = \|f - Q\|.$$

This  $Q$  is unique by Problem 4.3.3 if the norm  $\| \cdot \|_p$  is strictly convex, whereupon the claim follows. By Problem 4.3.4 this is the case for  $1 < p < \infty$ . In fact the norm is strictly convex even for  $p = 1$  over  $C[0, 1]$ , as is easily verified directly from the definition.

## Section 4.4 – Linear Transformations

### Problems

#### 4.4.2

Let  $T$  be an additive operator [that is,  $T(x_1 + x_2) = Tx_1 + Tx_2$ ] from a real normed linear space  $X$  into a normed linear space  $Y$ . If  $T$  is continuous, then  $T$  is homogeneous [that is,  $T(\lambda x) = \lambda Tx$ ]. [*Hint*: Prove that  $T[(m/n)x] = (m/n)Tx$ , where  $m, n$  are integers.]

*Solution.* Let  $x$  be an arbitrary element of  $X$ . By induction, additivity implies  $T(mx) = mTx$  for positive integers  $m$ . Moreover,

$$T0 = T(0 + 0) = T0 + T0$$

implies that  $T0 = 0$ , so that

$$0 = T0 = T(mx - mx) = T(mx) + T(-mx),$$

which shows that  $T(-mx) = -T(mx) = -mTx$ , thus extending the earlier result to nonpositive integers. Finally, if  $m$  and  $n$  are integers,  $n \neq 0$ , then

$$nT\left(\frac{m}{n}x\right) = T\left(n\frac{m}{n}x\right) = T(mx) = mTx,$$

so that  $T[(m/n)x] = (m/n)Tx$ , further extending the result to rationals.

Now, let  $\lambda \in \mathbb{R}$ . There exists a sequence  $(\lambda_n)$  in  $\mathbb{Q}$  such that  $\lambda_n \rightarrow \lambda$ . Clearly  $\lambda_n x \rightarrow \lambda x$  in  $X$ , so  $T(\lambda_n x) \rightarrow T(\lambda x)$  in  $Y$  by continuity of  $T$ . But, by what we found above,

$$T(\lambda_n x) = \lambda_n Tx \rightarrow \lambda Tx,$$

so  $T(\lambda x) = \lambda Tx$ .

#### 4.4.6

Find the norm of the operator  $A \in \mathcal{B}(X)$  given by  $(Af)(t) = tf(t)$  ( $0 \leq t \leq 1$ ), where (a)  $X = C[0, 1]$ , (b)  $X = L^p(0, 1)$  and  $(1 \leq p \leq \infty)$ .

*Solution.*

(a) For  $0 \leq t \leq 1$  we have  $|tf(t)| = |t||f(t)| \leq |f(t)|$ , hence  $\|Af\| \leq \|f\|$ , and

$$\|A\| = \sup_{f \neq 0} \frac{\|Af\|}{\|f\|} \leq 1.$$

The upper bound  $\|Af\|/\|f\| = 1$  is attained with  $f$  constant, so  $\|A\| = 1$ .

- (b) Let us first verify that  $A \in \mathcal{B}(X)$ , i.e. that it is a bounded linear operator  $L^p(0, 1) \rightarrow L^p(0, 1)$ . Linearity is immediate. If  $f \in \mathcal{L}^p(0, 1)$ , then  $|f|^p$  is integrable, so  $|tf(t)|^p = |t|^p|f(t)|^p$  is integrable by Corollary 2.10.2, and we see that  $A$  does indeed map into  $L^p(0, 1)$ . Finally, since  $|tf(t)|^p = |t|^p|f(t)|^p \leq |f(t)|^p$  for  $0 < t < 1$ , we have  $\|Af\|_p \leq \|f\|_p$ , so that  $\|A\| \leq 1$ .

We will now show that  $\|A\| \geq 1$ , so that  $\|A\| = 1$ . The case  $p = \infty$  is similar to (a), so we will assume  $1 \leq p < \infty$ . For  $n = 1, 2, \dots$ , define simple functions  $f_n : (0, 1) \rightarrow \mathbb{R}$  (or  $\mathbb{C}$ ) by

$$f_n(t) = \begin{cases} 0 & \text{if } 0 < t < 1 - 1/n, \\ n^{1/p} & \text{if } 1 - 1/n \leq t < 1. \end{cases}$$

Then one easily finds that  $\|f_n\|_p = 1$  and  $\|Af_n\|_p \geq 1 - 1/n$ , so that

$$\|A\| \geq \frac{\|Af_n\|_p}{\|f_n\|_p} \geq 1 - \frac{1}{n}$$

for all  $n$ . Indeed it follows that  $\|A\| \geq 1$ .

#### 4.4.7

A linear operator from a normed linear space  $X$  into a normed linear space  $Y$  is bounded if and only if it maps bounded sets onto bounded sets.

*Solution.* Let  $T : X \rightarrow Y$  be a linear operator between normed linear spaces.

Suppose first that  $T$  is bounded, and let  $A$  be a bounded subset of  $X$ . Then  $\|T\| < \infty$ , and there is some  $L$  such that  $\|x\| \leq L < \infty$  for all  $x \in A$ . Hence

$$\|Tx\| \leq \|T\| \|x\| \leq \|T\| L$$

for all  $x \in A$ , so  $T(A)$  is bounded.

Conversely, suppose that  $T$  maps bounded sets onto bounded sets. The set  $\{x \in X : \|x\| = 1\}$  is bounded, so there exists  $M$  such that  $\|Tx\| \leq M < \infty$  whenever  $\|x\| = 1$ . It follows that

$$\|T\| = \sup_{\|x\|=1} \|Tx\| \leq M,$$

and hence that  $T$  is bounded.

#### 4.4.8

A linear operator from a normed linear space  $X$  into a normed linear space  $Y$  is continuous if and only if it maps sequences converging to 0 into bounded sequences.

*Solution.* Let  $T : X \rightarrow Y$  be a linear operator between normed linear spaces. The claim is immediate if  $X$  is the trivial space (containing only the 0 vector), so let us assume that  $X$  is nontrivial.

Suppose first that  $T$  is continuous, and therefore bounded. Any sequence in  $X$  converging to 0 is easily seen to be bounded, hence is mapped to a bounded sequence by the conclusion of the previous problem.

Conversely, suppose  $T$  has the property that it maps sequences converging to 0 into bounded sequences. Assume towards a contradiction that  $T$  is unbounded. Then it is possible to construct a sequence  $(x_n)$  in  $X$  such that  $\|x_n\| = 1$  and  $\|Tx_n\| > n$  for every  $n$ . The sequence  $(x_n/\sqrt{n})$  converges to 0, so  $\{T(x_n/\sqrt{n})\}$  is bounded by hypothesis. But

$$\left\| T \left( \frac{x_n}{\sqrt{n}} \right) \right\| = \frac{1}{\sqrt{n}} \|Tx_n\| > \frac{1}{\sqrt{n}} \cdot n = \sqrt{n} \rightarrow \infty$$

as  $n \rightarrow \infty$ , yielding a contradiction. We conclude that  $T$  must be bounded, and thus continuous.

## Section 4.6 – The Open-Mapping Theorem and the Closed-Graph Theorem

### Problems

#### 4.6.1

If  $T, S, T^{-1}, S^{-1}$  belong to  $\mathcal{B}(X)$ , then  $(TS)^{-1} \in \mathcal{B}(X)$  and  $(TS)^{-1} = S^{-1}T^{-1}$ .

*Solution.* Note in particular that  $T, S, T^{-1}, S^{-1}$  belonging to  $\mathcal{B}(X)$  implies that all of these transformations are bijective, in addition to bounded. (If  $T$  is not surjective, then  $D_{T^{-1}} = T(X) \neq X$ , hence  $T^{-1} \notin \mathcal{B}(X)$ .)

It is clear that  $\|TS\| \leq \|T\| \|S\| < \infty$ , hence  $TS \in \mathcal{B}(X)$ , and it is bijective on account of being a composition of bijections. Thus  $(TS)^{-1}$  exists and is equal to  $S^{-1}T^{-1}$ . It follows that  $\|(TS)^{-1}\| \leq \|S^{-1}\| \|T^{-1}\| < \infty$ , hence  $(TS)^{-1} \in \mathcal{B}(X)$ .

#### 4.6.2

Let  $X$  be a Banach space and let  $A \in \mathcal{B}(X)$ ,  $\|A\| < 1$ . Prove that  $(I + A)^{-1}$  exists and is given by

$$(I + A)^{-1} = \sum_{n=0}^{\infty} (-1)^n A^n,$$

where the series is absolutely convergent [in  $\mathcal{B}(X)$ ]. Show also that

$$\|(I + A)^{-1}\| \leq 1/(1 - \|A\|).$$

*Solution.* We note first that

$$\sum_{n=0}^{\infty} \|(-1)^n A^n\| = \sum_{n=0}^{\infty} \|A^n\| \leq \sum_{n=0}^{\infty} \|A\|^n = \frac{1}{1 - \|A\|},$$



since  $\|A\| < 1$ . Hence the series  $\sum_{n=0}^{\infty} (-1)^n A^n$  is strongly convergent to an operator  $B \in \mathcal{B}(X)$ , by Theorem 4.5.2. (Recall that  $\mathcal{B}(X)$  is a Banach space, hence absolute convergence implies convergence. Convergence in  $\mathcal{B}(X)$  is the uniform convergence of operators, and uniform convergence implies strong convergence.)

For all  $x \in X$ , we have

$$ABx = A \left( \sum_{n=0}^{\infty} (-1)^n A^n x \right) = \sum_{n=0}^{\infty} (-1)^n A^{n+1} x = x - \sum_{n=0}^{\infty} (-1)^n A^n x = (I - B)x,$$

where we have used the continuity of  $A$  to interchange limiting processes. Also,

$$BAx = \sum_{n=0}^{\infty} (-1)^n A^{n+1} x = ABx.$$

It follows that

$$B(I + A) = (I + A)B = B + AB = B + I - B = I,$$

so that  $B = (I + A)^{-1}$ .

Finally,

$$\|(I + A)^{-1}\| \leq \sum_{n=0}^{\infty} \|(-1)^n A^n\| \leq \frac{1}{1 - \|A\|}$$

by what we found earlier.

### 4.6.3

Let  $X$  be a Banach space and let  $T$  and  $T^{-1}$  belong to  $\mathcal{B}(X)$ . Prove that if  $S \in \mathcal{B}(X)$  and  $\|S - T\| < 1/\|T^{-1}\|$ , then  $S^{-1}$  exists and is a bounded operator, and

$$\|S^{-1} - T^{-1}\| < \frac{\|T^{-1}\|}{1 - \|S - T\| \|T^{-1}\|}.$$

[Hint:  $S = [(S - T)T^{-1} + I]T$ .]

*Solution.* Note that  $ST^{-1} = I + (S - T)T^{-1}$ , and that

$$\|(S - T)T^{-1}\| \leq \|S - T\| \|T^{-1}\| < 1.$$

By the previous problem  $(ST^{-1})^{-1}$  exists, and

$$\|(ST^{-1})^{-1}\| \leq \frac{1}{1 - \|(S - T)T^{-1}\|} \leq \frac{1}{1 - \|S - T\| \|T^{-1}\|}.$$

Moreover, since  $ST^{-1}(ST^{-1})^{-1} = I$  and

$$T^{-1}(ST^{-1})^{-1}S = T^{-1}(ST^{-1})^{-1}ST^{-1}T = T^{-1}T = I,$$

we have  $S^{-1} = T^{-1}(ST^{-1})^{-1}$ . Finally,

$$\|S^{-1}\| = \|T^{-1}(ST^{-1})^{-1}\| \leq \|T^{-1}\| \|(ST^{-1})^{-1}\| \leq \frac{\|T^{-1}\|}{1 - \|(S - T)\| \|T^{-1}\|},$$

and

$$\|S^{-1} - T^{-1}\| = \|S^{-1}(S - T)T^{-1}\| \leq \|S^{-1}\| \|S - T\| \|T^{-1}\| < \|S^{-1}\|,$$

so that

$$\|S^{-1} - T^{-1}\| < \frac{\|T^{-1}\|}{1 - \|(S - T)\| \|T^{-1}\|}.$$

#### 4.6.4

Let  $X$  and  $Y$  be two linear vector spaces. Find necessary and sufficient conditions for a subset  $G$  of  $X \times Y$  to be the graph of a linear operator from  $X$  into  $Y$ .

*Solution.* We claim that  $G$  is the graph of a linear operator from  $X$  into  $Y$  if and only if

- (i)  $G$  is a linear subspace of  $X \times Y$ .
- (ii) The set  $G \cap (\{0\} \times Y)$  is a singleton.

It is clear that these conditions are necessary, so we need only prove that they are sufficient.

Since  $G$  is nonempty by (ii), it contains an element  $(x, y)$ . By (i) it also contains  $(0, 0) = 0 \cdot (x, y)$ , and it follows that  $G \cap (\{0\} \times Y) = \{(0, 0)\}$ .

If  $(u, v), (u, v') \in G$ , then

$$(0, v - v') = (u, v) - (u, v') \in G \cap (\{0\} \times Y)$$

by (i). It follows that  $(0, v - v') = (0, 0)$ , hence that  $v = v'$ .

By the above,  $G$  is a functional relation on  $X \times Y$ , so it defines a partial function  $T : D \subset X \rightarrow Y$ , where

$$D = \{x \in X; \exists y \in Y \text{ such that } (x, y) \in G\},$$

and  $T(x) = y$  if  $(x, y) \in G$ .

Suppose  $x_1, x_2 \in D$ , and  $\lambda_1, \lambda_2$  are scalars. Then

$$(\lambda_1 x_1 + \lambda_2 x_2, \lambda_1 T(x_1) + \lambda_2 T(x_2)) = \lambda_1 (x_1, T(x_1)) + \lambda_2 (x_2, T(x_2)) \in G$$

by (i). Hence  $\lambda_1 x_1 + \lambda_2 x_2 \in D$ , which shows that  $D$  is a linear subspace of  $X \times Y$ , and

$$T(\lambda_1 x_1 + \lambda_2 x_2) = \lambda_1 T(x_1) + \lambda_2 T(x_2),$$

so that  $T$  is a linear operator.

Finally, note that  $G$  is precisely the graph of  $T$ . This completes the proof.

#### 4.6.5

Let  $X$  and  $Y$  be Banach spaces and let  $T$  be a bounded linear map from  $X$  into  $Y$ . If  $T(X)$  is of the second category (in  $Y$ ), then  $T(X) = Y$ .

*Solution.* We will use the following lemma:

If  $W$  is a linear subspace of a normed space  $V$ , and if  $W$  contains a nonempty open subset of  $V$ , then  $W = V$ .

To see that this is true, note that  $W$  contains an open ball  $B(x_0, r)$ . Given any  $y \in V$ , let

$$x = \frac{r}{2\|y\|}y + x_0.$$

Then  $x \in B(x_0, r) \subset W$ , so that

$$y = \frac{2\|y\|}{r}(x - x_0) \in W$$

by closure under linear combinations.

Now, assume that  $T(X)$  is of the second category, and follow the steps of parts (a) and (b) of the proof of Theorem 4.6.1, but with  $T(X)$  in place of  $Y$ . We find that  $T(X)$  contains an open ball (see equation (4.6.2)) and hence that  $T(X) = Y$  by our lemma.

#### 4.6.6

Let  $X$  and  $Y$  be Banach spaces and let  $T$  be a linear map from a linear subspace  $D_T$  of  $X$  into  $Y$ . If  $D_T$  (in  $X$ ) and the graph of  $T$  (in  $X \times Y$ ) are closed, then  $T$  is bounded—that is,  $\|Tx\| \leq K\|x\|$  for all  $x \in D_T$  ( $K$  constant).

*Solution.* Note first that  $D_T$  is complete; this is true for any closed subset of any complete space. Indeed, if  $(x_n)$  is a Cauchy sequence in  $D_T$ , then it is also a Cauchy sequence in  $X$ , so it converges to a point  $x \in X$ , and since  $D_T$  is closed we have  $x \in D_T$ . It follows that  $D_T$  is a Banach space. Hence we can regard  $T$  as a map  $D_T \rightarrow Y$  and apply Theorem 4.6.4 to conclude that  $T$  is continuous, thus bounded by Theorem 4.4.2.

#### 4.6.7

Let  $X$  be a normed linear space with any one of two norms  $\|\cdot\|_1, \|\cdot\|_2$ . If  $\|x_n\|_2 \rightarrow 0$  implies  $\|x_n\|_1 \rightarrow 0$ , then (4.6.5) holds.

*Solution.* Assume towards a contradiction that (4.6.5) does not hold. Then, for every  $n \in \{1, 2, \dots\}$ , there exists  $x_n \in X$  such that  $\|x_n\|_1 > n\|x_n\|_2$ . The sequence  $y_n = x_n/(n\|x_n\|_2)$  satisfies

$$\|y_n\|_1 = \frac{\|x_n\|_1}{n\|x_n\|_2} > \frac{n\|x_n\|_2}{n\|x_n\|_2} = 1$$

for all  $n$ . But  $\|y_n\|_2 = 1/n \rightarrow 0$ , which by hypothesis implies  $\|y_n\|_1 \rightarrow 0$ , yielding a contradiction.

## Section 4.8 – The Hahn-Banach Theorem

### Problems

#### 4.8.1

Let  $X$  be a normed linear space and let  $\{x_n\} \subset X$ . A point  $y_0$  is the limit of linear combinations  $\sum_{j=1}^n c_j x_j$  if and only if  $x^*(y_0) = 0$  for all  $x^*$  for which  $x^*(x_j) = 0$  for  $1 \leq j < \infty$ .

*Solution.* Note that “ $y_0$  is the limit of linear combinations  $\sum_{j=1}^n c_j x_j$ ” is not meant to imply  $y_0 = \lim_{n \rightarrow \infty} \sum_{j=1}^n c_j x_j$  for some sequence of coefficients  $(c_j)$ . Rather, it just says that  $y_0$  is the limit of a sequence of finite linear combinations of elements in  $\{x_n\}$ ; i.e., that  $y_0$  is a limit point of  $S := \text{span}\{x_n\}$ . Also, the statement “ $x^*(x_j) = 0$  for  $1 \leq j < \infty$ ” can be simplified to “ $x^*$  vanishes on  $S$ .”

Suppose that  $y_0$  is a limit point of  $S$ . Then there is a sequence  $(s_n)$  in  $S$  such that  $s_n \rightarrow y_0$ . If  $x^* \in X^*$  vanishes on  $S$ , then

$$x^*(y_0) = x^*\left(\lim_{n \rightarrow \infty} s_n\right) = \lim_{n \rightarrow \infty} x^*(s_n) = 0$$

by continuity of  $x^*$ .

Conversely, suppose that  $x^*(y_0) = 0$  for all  $x^* \in X^*$  that vanish on  $S$ . Let

$$d = \inf_{s \in S} \|s - y_0\|.$$

If  $d > 0$ , then Theorem 4.8.3 tells us that there exists  $x^* \in X^*$  such that  $x^*(y_0) = 1$ , but which vanishes on  $S$ . This contradicts our assumptions, so we conclude that  $d = 0$ , which in turn shows that  $y_0$  is a limit point of  $S$ .

#### 4.8.5

Let  $X$  be an infinite-dimensional Banach space. Prove that there exists an infinite, strictly decreasing sequence  $\{Y_n\}$  of infinite-dimensional closed linear subspaces of  $X$ . [*Hint:* Take  $Y_1$  to be the null space of some  $x_1^* \neq 0$  in  $X^*$ . Take  $Y_2$  to be the null space of some  $x_2^* \neq 0$  in  $Y_1^*$ , and so on.]

*Solution.* We will construct such a sequence by induction. For the base case, let  $Y_0 = X$ . Certainly  $Y_0$  is an infinite-dimensional closed linear subspace of  $X$ .

For the inductive step, assume that we have infinite-dimensional closed linear subspaces  $Y_0 \supset Y_1 \supset \cdots \supset Y_n$ , with the inclusions being strict. Fix an element  $x_0 \in Y_n$  such that  $\|x_0\| = 1$ . (Such an element is guaranteed to exist since  $Y_n$  is infinite-dimensional.) Corollary 4.8.4 provides a continuous linear functional  $x^* \in Y_n^*$  such that  $x^*(x_0) = 1$ . Let  $Y_{n+1}$  be the null space of  $x^*$ ; clearly a proper linear subspace of  $Y_n$ , hence also of  $X$ . As discussed after Corollary 4.8.7, each element  $x \in Y_n$  can be written as  $x = z + \lambda x_0$ , where  $\lambda = x^*(x)$  and  $z = x - \lambda x_0 \in Y_{n+1}$ . It follows that  $Y_{n+1}$  is infinite-dimensional. Finally, if  $(y_i)$  is a sequence in  $Y_{n+1}$  and  $y_i \rightarrow y \in X$ , then  $x^*(y) = \lim_{i \rightarrow \infty} x^*(y_i) = 0$  by continuity, so  $y \in Y_{n+1}$ , making  $Y_{n+1}$  a closed subset of  $X$ . This concludes the inductive step.

#### 4.8.9

Let  $u(t)$  be a function defined on  $a < t < b$  with values in a Banach space  $X$ . We say that  $u(t)$  is *strongly differentiable at  $t$  [on  $(a, b)$ ]* if  $\lim_{h \rightarrow 0} \{[u(t+h) - u(t)]/h\}$  exists [for all  $t \in (a, b)$ ]. The limit is denoted by  $du(t)/dt$  and is called the derivative of  $u(t)$ . For functions  $A(t)$  with values in  $\mathcal{B}(X)$ , if  $\lim_{h \rightarrow 0} \{[A(t+h)x - A(t)x]/h\}$  exists for any  $x \in X$ , then we say that  $A(t)$  has a *strong derivative*. If  $\lim_{h \rightarrow 0} \{[A(t+h) - A(t)]/h\}$  exists (in the uniform topology), then we say that  $A(t)$  is *uniformly differentiable*. Prove that  $e^{tA}$  [ $A \in \mathcal{B}(X)$ ] is uniformly differentiable and  $de^{tA}/dt = Ae^{tA}$ .

*Solution.* First note that, given any  $t \in \mathbb{R}$ ,

$$\sum_{n=0}^{\infty} \left\| \frac{(tA)^n}{n!} \right\| \leq \sum_{n=0}^{\infty} \frac{(|t|^n \|A\|^n)}{n!} = e^{|t| \|A\|} < \infty,$$

so that the series  $\sum_{n=0}^{\infty} (tA)^n/(n!)$  is strongly convergent (to an operator) in  $\mathcal{B}(X)$  (c.f. Theorems 4.1.2, 4.5.2). Hence we can define a map  $E_A : \mathbb{R} \rightarrow \mathcal{B}(X)$  by  $E_A(t) = \sum_{n=0}^{\infty} (tA)^n/(n!)$ . Instead of  $E_A(t)$  we typically write  $e^{tA}$ .

With a little work we find that

$$\frac{e^{(t+h)A} - e^{tA}}{h} - Ae^{tA} = \sum_{n=2}^{\infty} \frac{1}{n!} \frac{(t+h)^n - t^n - nht^{n-1}}{h} A^n.$$

Further simplification with the binomial formula yields

$$\frac{(t+h)^n - t^n - nht^{n-1}}{h} = \sum_{k=2}^n \binom{n}{k} t^{n-k} h^{k-1}.$$

Hence

$$\left\| \frac{e^{(t+h)A} - e^{tA}}{h} - Ae^{tA} \right\| \leq \sum_{n=2}^{\infty} \frac{\|A\|^n}{n!} \sum_{k=2}^n \binom{n}{k} |t|^{n-k} |h|^{k-1}.$$

If  $t = 0$ , then the right-hand side becomes

$$|h| \cdot \|A\|^2 \sum_{n=0}^{\infty} \frac{(|h| \|A\|)^n}{(n+2)!} \leq |h| \|A\|^2 e^{|h| \|A\|},$$

which goes to zero as  $h \rightarrow 0$ . If  $t \neq 0$ , then

$$\sum_{k=2}^n \binom{n}{k} |t|^{n-k} |h|^{k-1} \leq |h| |t|^{n-2} \sum_{k=2}^n \binom{n}{k} = |h| |t|^{n-2} 2^n$$

whenever  $|h| \leq |t|$ , yielding

$$\left\| \frac{e^{(t+h)A} - e^{tA}}{h} - Ae^{tA} \right\| \leq \frac{|h|}{|t|^2} \sum_{n=2}^{\infty} \frac{(2|t| \|A\|)^n}{n!} \leq \frac{|h|}{|t|^2} e^{2|t| \|A\|}$$

for sufficiently small  $h$ , which also goes to zero as  $h \rightarrow 0$ . It follows that  $e^{tA}$  is uniformly differentiable on  $\mathbb{R}$ , with derivative  $Ae^{tA}$ .

#### 4.8.10

Let  $X$  be a real normed linear space, and let  $u(t)$  be continuous and strongly differentiable in  $(a, b)$ . Then for any  $a < \alpha < \beta < b$ ,

$$\|u(\beta) - u(\alpha)\| \leq (\beta - \alpha) \sup_{\alpha \leq t \leq \beta} \left\| \frac{du(t)}{dt} \right\|.$$

[Hint: Apply  $x^*$  to  $u(\beta) - u(\alpha)$ .]

*Solution.* If  $u(\alpha) = u(\beta)$  then there is nothing to prove, so assume  $u(\beta) \neq u(\alpha)$ . By Corollary 4.8.4 there is a bounded linear operator  $x^* \in X^*$  such that  $\|x^*\| = 1$  and  $x^*(u(\beta) - u(\alpha)) = \|u(\beta) - u(\alpha)\|$ . Define  $f = x^* \circ u : (a, b) \rightarrow \mathbb{R}$ . Since  $x^*$  is linear and continuous,

$$\lim_{h \rightarrow 0} \frac{f(t+h) - f(t)}{h} = \lim_{h \rightarrow 0} x^* \left[ \frac{u(t+h) - u(t)}{h} \right] = x^* \left[ \lim_{h \rightarrow 0} \frac{u(t+h) - u(t)}{h} \right],$$

so that  $f$  is differentiable with derivative

$$\frac{df(t)}{dt} = x^* \left[ \frac{du(t)}{dt} \right].$$

By the mean value theorem from elementary real analysis, we have

$$f(\beta) - f(\alpha) = (\beta - \alpha) \frac{df(\gamma)}{dt}$$

for some  $\gamma \in (\alpha, \beta)$ . Thus,

$$\begin{aligned} \|u(\beta) - u(\alpha)\| &= x^*(u(\beta) - u(\alpha)) \\ &= f(\beta) - f(\alpha) \\ &= (\beta - \alpha) \frac{df(\gamma)}{dt} \\ &= (\beta - \alpha) x^* \left[ \frac{du(\gamma)}{dt} \right] \\ &\leq (\beta - \alpha) \|x^*\| \cdot \left\| \frac{du(\gamma)}{dt} \right\| \\ &\leq (\beta - \alpha) \sup_{\alpha \leq t \leq \beta} \left\| \frac{du(t)}{dt} \right\|. \end{aligned}$$

#### 4.8.12

For every normed linear space  $X$  there is a set  $A$  such that  $X$  is isomorphic to a subspace of the Banach space of functions  $f$  on  $A$  with norm  $\|f\| = \sup_{t \in A} |f(t)|$ . If  $X$  is separable,  $A$  is countable. [Hint: Let  $\{x_\alpha : \alpha \in A\}$  be dense in  $X$ . Let  $f(x, \alpha)$  be the bounded linear functional (in  $x \in X$ ) satisfying  $\|f(\cdot, \alpha)\| = 1$ ,  $f(x_\alpha, \alpha) = \|x_\alpha\|$ . Define the isomorphism  $x \rightarrow g_x(\alpha)$ , where  $g_x(\alpha) = f(x, \alpha)$ . Prove:  $\|f(x, \alpha) - \|x\|\| \leq 2\|x_\alpha - x\|$ .]

*Solution.* The conclusion is immediate if  $X$  is a trivial space, so assume  $X \neq \{0\}$ . Let  $\{x_\alpha\}_{\alpha \in A}$  be any dense subset of  $X$  (possibly  $X$  itself), indexed by a set  $A$ . (Recall that any space can be indexed by itself.) For each  $\alpha \in A$ , Corollary 4.8.4 fashions a bounded linear functional  $f_\alpha \in X^*$  such that  $\|f_\alpha\| = 1$  and  $f_\alpha(x_\alpha) = \|x_\alpha\|$ .

For each  $x \in X$ , define  $g_x : A \rightarrow \mathbb{F}$  (where  $\mathbb{F}$  is the field associated with  $X$ ) by  $g_x(\alpha) = f_\alpha(x)$ . We will prove the following properties of the functions  $g_x$ :

- (i) If  $x, y \in X$  and  $\lambda, \mu \in \mathbb{F}$ , then  $\lambda g_x + \mu g_y = g_{\lambda x + \mu y}$ .
- (ii) Each function  $g_x$  is bounded, with  $\sup_{\alpha \in A} |g_x(\alpha)| = \|x\|$ .

To prove (i), simply let  $\alpha \in A$  and compute

$$(\lambda g_x + \mu g_y)(\alpha) = \lambda g_x(\alpha) + \mu g_y(\alpha) = \lambda f_\alpha(x) + \mu f_\alpha(y) = f_\alpha(\lambda x + \mu y) = g_{\lambda x + \mu y}(\alpha),$$

using the linearity of  $f_\alpha$ .

Proving (ii) takes more work. Note first that

$$\sup_{\alpha \in A} |g_x(\alpha)| = \sup_{\alpha \in A} |f_\alpha(x)| \leq \sup_{\alpha \in A} \|f_\alpha\| \|x\| = \|x\|.$$

Next, given any  $\epsilon > 0$ , fix  $\alpha' \in A$  such that  $\|x_{\alpha'} - x\| < \epsilon/2$ . (This is possible because  $\{x_\alpha\}_{\alpha \in A}$  is dense in  $X$ .) By the triangle inequality,

$$\|x\| \leq \|x\| - \|x_{\alpha'}\| + \|x_{\alpha'}\| - |f_{\alpha'}(x)| + |f_{\alpha'}(x)|.$$

Now, the “reverse triangle inequality” yields

$$\|x\| - \|x_{\alpha'}\| \leq \|x_{\alpha'} - x\| < \frac{\epsilon}{2}$$

and

$$\begin{aligned} \left| \|x_{\alpha'}\| - |f_{\alpha'}(x)| \right| &\leq \left| \|x_{\alpha'}\| - f_{\alpha'}(x) \right| \\ &= |f_{\alpha'}(x_{\alpha'}) - f_{\alpha'}(x)| \\ &= |f_{\alpha'}(x_{\alpha'} - x)| \\ &\leq \|f_{\alpha'}\| \|x_{\alpha'} - x\| \\ &= \|x_{\alpha'} - x\| \\ &< \frac{\epsilon}{2}. \end{aligned}$$

Also,

$$|f_{\alpha'}(x)| \leq \sup_{\alpha \in A} |f_\alpha(x)| = \sup_{\alpha \in A} |g_x(\alpha)|.$$

Putting it all together, we have

$$\|x\| \leq \sup_{\alpha \in A} |g_x(\alpha)| + \epsilon,$$

and since this holds for every  $\epsilon$  we finally arrive at (ii).

Let  $F_A$  be the Banach space of all bounded functions  $A \rightarrow \mathbb{F}$ , with the supremum norm  $\|g\| = \sup_{\alpha \in A} |g(\alpha)|$ . Since the functions  $g_x$  are bounded by (ii), we can define a map  $\sigma : X \rightarrow F_A$  by  $\sigma(x) = g_x$ . The image of  $\sigma$  is a linear subspace of  $F_A$  by (i). Moreover,  $\sigma$  is an imbedding, since

$$\|\sigma(x) - \sigma(y)\| = \|g_x - g_y\| = \|g_{x-y}\| = \sup_{\alpha \in A} |g_{x-y}(\alpha)| = \|x - y\|$$

for all  $x, y \in X$ , by (i) and (ii). Every imbedding is injective, so  $\sigma$  is an isomorphism (in the sense of Section 3.3) onto its image.

Finally, if  $X$  is separable, then we can take  $A$  to be countable.



# Chapter 6 – Hilbert Spaces and Spectral Theory

## Section 6.2 – The Projection Theorem

### Problems

#### 6.2.1

If  $M$  and  $N$  are closed linear spaces and  $M \perp N$ , then  $M \oplus N$  is a closed linear space.

*Solution.* We start by noting that an analogue of the Pythagorean theorem holds in Hilbert spaces, and in fact more generally in inner product spaces. Namely, if  $u \perp v$  then

$$\|u + v\|^2 = (u + v, u + v) = (u, u) + (u, v) + (v, u) + (v, v) = \|u\|^2 + \|v\|^2.$$

We will assume that  $M, N \subset H$ , with  $H$  a Hilbert space. It is clear that  $M \oplus N$  is a linear subspace, so it remains only to show that it is closed. Let  $(x_n)$  be a sequence in  $M \oplus N$  with limit  $x \in H$ . For each  $n$  we have  $x_n = y_n + z_n$  for some  $y_n \in M$  and  $z_n \in N$ , and  $x = y + z$  for some  $y \in M$  and  $z \in M^\perp$  by the projection theorem (Theorem 6.2.2). Note that  $y_n - y \in M$  and  $z_n - z \in M^\perp$ , so that  $(y_n - y) \perp (z_n - z)$ . Thus

$$\|z_n - z\| \leq \|y_n - y + z_n - z\| = \|x_n - x\|$$

by the analogue of the Pythagorean theorem with  $u = y_n - y$  and  $v = z_n - z$ . It follows that  $z_n \rightarrow z$ , so that  $z \in N$  since  $N$  is closed. This in turn means that  $x \in M \oplus N$ , hence that  $M \oplus N$  is closed.

#### 6.2.2

Let  $M$  be any subset of a Hilbert space  $H$ . Then  $(M^\perp)^\perp$  is the closed linear space spanned by  $M$ .

*Solution.* We begin with a small lemma:

If  $C$  is a closed linear subspace of a Hilbert space, then  $(C^\perp)^\perp = C$ .

Indeed, it is clear that  $C \subset (C^\perp)^\perp$ . Conversely, suppose  $x \in (C^\perp)^\perp$ . By the projection theorem  $x = y + z$  for some  $y \in C$  and  $z \in C^\perp$ . But

$$\|z\|^2 = (z, z) = (x - y, z) = (x, z) - (y, z) = 0$$

by orthogonality, so  $z = 0$  and  $x = y \in C$ . Hence  $(C^\perp)^\perp \subset C$ , and the lemma follows.

Note that “the closed linear space spanned by  $M$ ” is the intersection of all linear subspaces of  $H$  that contain  $M$  (by definition; see Section 4.2). Let us denote this subspace by  $C$ . We immediately see that  $C \subset (M^\perp)^\perp$ , since the latter is a closed linear subspace containing  $M$ . Moreover, since  $M \subset C$ , we have  $C^\perp \subset M^\perp$ . By taking orthogonal complements once more and applying the lemma, we obtain

$$(M^\perp)^\perp \subset (C^\perp)^\perp = C,$$

thus proving that  $(M^\perp)^\perp = C$ .